In the last lecture we showed that for the class of acyclic Conjunctive Queries, evaluation is in polynomial time (combined complexity). It is logical to ask whether acyclic queries is the only class of queries that can be evaluated in polynomial time. The answer to this question is that there are other classes of CQs where polynomial time evaluation is possible as well.

Example 1

Consider the Boolean cycle query, which asks whether there is cycle of length $k$:

$$C^k() \leftarrow R_1(x_1, x_2), R_2(x_2, x_3), \ldots, R_k(x_k, x_1)$$

Assume that for every $i$, $|R_i| = N$. As we showed in the last lecture, this query is not acyclic and does not admit a join tree. However, we can apply a similar idea to evaluate this query in polynomial time. We first perform the natural join $R(x_1, x_2) \bowtie R(x_2, x_3)$, then project on $x_1, x_3$, then join with $R(x_3, x_4)$, project on $x_1, x_4$, and so on. The key difference from the algorithm that evaluates the path query $P^k$ is that we also keep $x_1$ around. Observe that at every point the size of the intermediate result is at most $N^2$. Hence, we can implement the algorithm with running time time $O(kN^2)$, and the running time is again polynomial in the size of the query and the input!

The characterization of CQs that we can compute in polynomial time is based on the idea of "decomposing" the hypergraph of the query. We start by presenting one such decomposition, called the query decomposition. We can think of the query decomposition as a generalization of the construct of a join forest for acyclic queries.

5.1 Tree Decompositions

We start by providing the definition of a tree decomposition.
Definition 1: Tree Decomposition

A tree decomposition of a CQ \( q \) is a pair \( (T, \chi) \), where (i) \( T = (V, E) \) is a tree, and (ii) \( \chi \) is a labeling function which associates to each vertex \( v \in V \) a subset of variables of \( q \), \( \chi(v) \), such that the following conditions are satisfied:

1. For each atom \( A \) in the body of \( q \), there exists a node \( v \in V \) such that \( \chi(v) \) contains all variables of \( A \).
2. For each variable \( x \in \text{vars}(q) \), the set of nodes \( \{ v \mid x \in \chi(v) \} \) forms a connected subtree.

The set \( \chi(v) \) is also commonly called a bag. The second condition of the decomposition generalizes the "connectedness" condition for join trees and is equivalent to requiring that for every two vertices \( v_1, v_2 \) of the tree, the variables in \( \chi(v_1) \cap \chi(v_2) \) will appear in all the vertices along the unique path that connects \( v_1 \) to \( v_2 \). When each bag of a tree decomposition consists of the variables of a single relation, it is easy to see that the decomposition is exactly a join tree.

Given a node \( v \) in the tree decomposition, we define its width \( w(v) \) to be the minimum number of atoms necessary to cover all variables in \( \chi(v) \) (meaning that every variable must belong in some atom).

Definition 2: Generalized Hypertree Width

Let \( (T, \chi) \) be a tree decomposition of a CQ \( q \). The generalized hypertree width (ghw) of the decomposition is defined as \( \max_{v \in V} w(v) \). The ghw of a query \( q \) is the minimum ghw over all possible tree decompositions.

To give an example, consider the cycle query \( C_k \) of the initial example. We will construct three different decompositions for this query:

1. The first decomposition is a tree with a single node \( v \), and \( \chi(v) = \{ x_1, x_2, \ldots, x_k \} \). It is trivial to see that this decomposition satisfies both conditions. To compute the width of the node \( v \), observe that if we choose every second atom (i.e., \( R_1, R_3, R_5, \ldots \)), we can cover all variables in \( \chi(v) \). Hence, the width of \( v \) is \( \lceil k/2 \rceil \), and the ghw of the decomposition is \( \lceil k/2 \rceil \).
2. The second decomposition is a tree with two nodes \( v_1, v_2 \), and an edge \( \{ v_1, v_2 \} \). The labeling function is \( \chi(v_1) = \{ x_1, x_2, \ldots, x_{k-1} \}, \chi(v_2) = \{ x_{k-1}, x_k, x_1 \} \). This is a correct decomposition because \( v_1, v_2 \) share 2 variables and indeed there are connected. The ghw in this case is \( \max\{ \lceil k/2 \rceil, 2 \} = \lceil k/2 \rceil \) again.
3. The third decomposition has \( k \) vertices \( v_1, \ldots, v_k \) that form a path, where \( \chi(v_i) = \{ x_1, x_i, x_{i+1} \} \). It is easy to see that the first condition of a tree decomposition is satisfied. For the second condition, notice that every variable apart from \( x_1 \) appears in consecutive vertices in the path that satisfies the connectedness property. Variable \( x_1 \) appears in every vertex, so it trivially satisfies the property. The ghw here is 2, since each bag can be covered by two atoms.
Can we construct a tree decomposition for $C_k$ with ghw less than 2? The lemma below tells us that it is not possible to do, which shows that the ghw of $C_k$ is 2.

**Proposition 1**

A CQ $q$ is acyclic if and only if $ghw(q) = 1$.

Tree decompositions can be interpreted as query plans for joins. Given a decomposition, we can first compute one intermediate relation per bag/node, by computing the join restricted to the variables of the bag. Formally, given a bag with variables $U \subseteq \text{vars}(q)$, we define the query $q[U]$ as one obtained by first projecting each atom $A$ on $U \cap \text{vars}(A)$, and then computing the natural join. For instance, if $U = \{x_1, x_2, x_3\}$, the query $C_k[U]$ computes the natural join $\pi_{x_1}(R_k) \bowtie R_1 \bowtie R_2 \bowtie \pi_{x_3}(R_3)$. The intermediate relations can then be joined using Yannakakis algorithm.

**Algorithm 1: Evaluation of CQ with ghw $k$**

<table>
<thead>
<tr>
<th>Input: Conjunctive Query $q$ with $ghw = k$, instance $I$</th>
<th>Output: $q(I)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>construct a tree decomposition $(T, \chi)$ for $q$ ;</td>
<td></td>
</tr>
<tr>
<td>for vertex $v$ in $T$ do</td>
<td></td>
</tr>
<tr>
<td></td>
<td>compute $q<a href="I">\chi(v)</a>$ ;</td>
</tr>
<tr>
<td>end</td>
<td></td>
</tr>
<tr>
<td>Run Yannakakis using $T$ as the join tree and $q<a href="I">\chi(v)</a>$ as the input relations of each node ;</td>
<td></td>
</tr>
</tbody>
</table>

In the above algorithm, the key observation is that we need $O(|I|^k)$ time to compute the intermediate result for each node in the decomposition. Moreover, the size of each intermediate result is bounded by $O(|I|^k)$. Hence, applying Yannakakis algorithm leads to the following results [CR00]:

**Theorem 1**

Let $q$ be a boolean CQ with $ghw(q) = k$. Given a tree decomposition $q$ of width $k$, we can compute $q$ with running time $O(|I|^k)$, where $I$ is the input database.

**Theorem 2**

Let $q$ be a full CQ with $ghw(q) = k$. Given a tree decomposition of $q$ of width $k$, we can compute $q$ with running time $O(|I|^k + \text{OUT})$.

This implies that we can efficiently evaluate a Conjunctive Query with bounded ghw, where bounded means that the width is bounded by some constant. Unlike the case of acyclic CQs, no efficient method for checking whether the ghw is bounded is known. In fact, deciding whether a CQ has a decomposition with bounded width is NP-complete.
5.2 Other Notions of Width

Apart from general hypertree width, several other notions of decompositions have been presented in the literature:

- **Hypertree-Width**: Bounded hypertree-width implies evaluation in polynomial time [GSL02]. The hypertree-width is always larger than the generalized hypertree width, but we can always find a bounded hypertree decomposition, if one exists, in polynomial time.

- **Fractional Hypertree-Width**: This is a generalization of the hypertree-width that encompasses an even bigger class of CQs that can be evaluated in polynomial time. We will talk more about these in subsequent lectures!

- **Submodular Width**: This is the currently best known width measure for a CQ.

We should note here that decompositions and the evaluation of CQs is very tightly connected with the area of Constraint Satisfaction Problems (CSPs) in artificial intelligence; see the survey [GGS] for more details on this connection and the applications of tree decompositions. In fact, a tree decomposition is essentially the same notion as the *junction tree* used used in inference for graphical models.

References


