# Join Queries with Negation (and Aggregation)

**Conjunctive Queries with Negation and Aggregation: A Linear Time Characterization [PODS'24]** Hangdong Zhao, Austen Fan, Xiating Ouyang, Paraschos Koutris





#### This Talk

Much progress has been over the last years on faster join algorithms

- worst-case optimal joins
- constant-delay enumeration
- tree decompositions & width measures
- PANDA

What happens when we add negation (and aggregation)?

# **Conjunctive Queries (CQs)**

head



- variables  $\mathbf{x} = \{x_1, ..., x_n\}$
- hypergraph ( $[n], \mathscr{E}$ )
- for a hyperedge  $E \subseteq [n]$  :  $\mathbf{x}_E = \{x_i\}_{i \in E}$

body  $Q(\mathbf{x}_F) = \bigwedge R_K(\mathbf{x}_K)$  $K \in \mathscr{E}$ 

- Boolean:  $F = \emptyset$
- full:  $F = [n] = \{1, 2, ..., n\}$

# Example: Triangle

#### $Q(x_1, x_3) = R(x_1, x_2) \land S(x_2, x_3) \land T(x_1, x_3)$



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# **Conjunctive Queries + Negation (CQNs)**

head 
$$points$$
  
 $Q(\mathbf{x}_F) = \bigwedge_{K \in \mathscr{C}^+} R_F$ 

• We need a **safety** condition: the positive atoms must contain all variables • The hypergraph ( $[n], \mathscr{E}^+, \mathscr{E}^-$ ) is called the **signed hypergraph** 



# Example: Open Triangle

#### $Q(x_2) = R(x_1, x_2) \land S(x_2, x_3) \land \neg T(x_1, x_3)$





### Example: 3-independent set

#### $Q() = V(x_1) \wedge V(x_2) \wedge V(x_3) \wedge \neg R(x_3) \wedge \nabla R(x_3)$

 $Q() = \neg R(x_1, x_2) \land \neg R(x_2, x_3) \land \neg R(x_1, x_3)$ 

$$(x_1, x_2) \land \neg R(x_2, x_3) \land \neg R(x_1, x_3)$$

If all positive relations are singleton, we will sometimes ignore them and just write



Some Background

# α-acyclicity

#### $Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_4, x_5) \land U(x_3, x_6)$

#### A CQ is $\alpha$ -acyclic if and only if it admits a join tree



### The structure of $\alpha$ -acyclicity

A node *v* is an  $\alpha$ -leaf if the set { $K \in \mathscr{C}$  |  $v \in K$ } contains a maximum element (pivot)

 $x_1: R(x_1, x_2) \wedge S(x_2, x_3) \wedge T(x_3, x_1) \wedge U(x_1, x_2, x_3)$  $x_2: R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land U(x_1, x_2, x_3)$  $x_3: R(x_1, x_2) \wedge S(x_2, x_3) \wedge T(x_3, x_1) \wedge U(x_1, x_2, x_3)$ 

All variables are  $\alpha$ -leaves for this hypergraph!

#### $Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land U(x_1, x_2, x_3)$

# The structure of $\alpha$ -acyclicity

- A CQ is  $\alpha$ -acyclic **iff** it admits an  $\alpha$ -elimination sequence. At every step: find any  $\alpha$ -leaf *x* (with pivot *R*) 1.
- remove any relation with variables contained in *R* 2.
- remove *x* from *R* 3.

$$Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land U(x_1, x_2, x_3)$$

 $x_2$  is an  $\alpha$ -leaf :  $S(x_2, x_3) \wedge U(x_2, x_3)$  $x_3$  is an  $\alpha$ -leaf :  $U(x_3)$ 

 $x_1$  is an  $\alpha$ -leaf :  $R(x_1, x_2) \wedge S(x_2, x_3) \wedge T(x_3, x_1) \wedge U(x_1, x_2, x_3)$ 

### A linear-time algorithm

We follow the  $\alpha$ -elimination sequence. At every step:

- find any  $\alpha$ -leaf *x* (with pivot *R*) 1.
- for any T with variables contained in R, update  $R \leftarrow R \ltimes T$  and remove T 2.
- project out *x* from *R* 3.

### A linear-time algorithm

 $Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land U(x_1, x_2, x_3)$ 

 $x_1$  is an  $\alpha$ -leaf :  $R(x_1, x_2) \wedge S(x_2, x_3) \wedge T(x_3, x_1) \wedge U(x_1, x_2, x_3)$  $x_2$  is an  $\alpha$ -leaf :  $S(x_2, x_3) \wedge U(x_2, x_3)$  $x_3$  is an  $\alpha$ -leaf :  $U(x_3)$ 



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### A linear-time characterization for CQs

[Yannakakis '81] For an  $\alpha$ -acyclic CQ with input size N:

- if it is **Boolean**, it can be evaluated in linear time O(N)1.
- if it is **full**, the output can be enumerated with constant delay after 2. linear-time preprocessing, with total time O(N + OUT)
- 3. if it is **full**, we can count the answers in linear time O(N)

believed conjectures

Moreover, no other CQs admit linear-time algorithms under widely

What is the linear-time characterization for CQs with negation?

### The Inclusion-Exclusion Principle

We can rewrite this query using a **difference** operator:

$$Q = \left( R(x_1, x_2) \land S(x_2, x_3) \right)$$

$$Q1: acyclic CQ$$

 $#Q = #Q_1 - #Q_2$ 

 $Q(x_1, x_2, x_3) = R(x_1, x_2) \wedge S(x_2, x_3) \wedge \neg T(x_1, x_2, x_3)$ 

 $-(R(x_1, x_2) \land S(x_2, x_3) \land T(x_1, x_2, x_3))$ Q2: acyclic CQ

## The Inclusion-Exclusion Principle

$$Q(\mathbf{x}) = \bigwedge_{K \in \mathscr{C}^+} R_K$$

We can generalize this idea via the inclusion-exclusion principle [Brault-Baron '13]:

$$#Q = \sum_{S \subseteq \mathscr{E}}$$

where  $Q_S$  is the CQ with hypergraph ([n],  $\mathscr{E}^+ \cup S$ )

 $K_K(\mathbf{x}_K) \wedge \bigwedge \neg R_K(\mathbf{x}_K)$  $K \in \mathscr{E}^{-}$ 

 $(-1)^{|S|} # Q_S$ 

# Signed-acyclicity

be evaluated in linear time (data complexity)

- Caveat #1: the algorithm is *exponential* in the size of the query
- Caveat #2: we cannot use this idea to perform *constant-delay enumeration*

If the hypergraph  $\mathscr{E}^+ \cup S$  is  $\alpha$ -acyclic for any  $S \subseteq \mathscr{E}^-$  then #Q (and thus Boolean Q) can

#### A CQ with negation is **signed-acyclic** if $\mathscr{E}^+ \cup S$ is $\alpha$ -acyclic for any $S \subseteq \mathscr{E}^-$

Brault-Baron '13]

### Signed-acyclicity: examples

#### 

 $Q() = \neg R(x_1, x_2) \land \neg S(x_2, x_3) \land \neg T(x_3, x_4)$ 

 $Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land U(x_1, x_2, x_3)$ 

 $Q() = R(x_1, x_2) \land S(x_2, x_3) \land T(x_3, x_1) \land \neg U(x_1, x_2, x_3)$ 

 $Q() = \neg R(x_1, x_2) \land \neg S(x_2, x_3) \land \neg T(x_3, x_1) \land \neg U(x_1, x_2, x_3)$ 

# β-acyclicity

- Suppose all positive relations are unary (arity = 1)
- Then signed-acyclicity is equivalent to: any subset of  $\mathscr{E}^-$  is  $\alpha$ -acyclic
- This is equivalent to the notion of β-acyclicity [Duris '12, Brault-Baron '14]
- Existing algorithms for β-acyclic CQNs include polylogarithmic factors

A Linear-Time Algorithm

# The structure of signed-acyclicity

- A node v is a **signed-leaf** if there exists  $K \in \mathscr{C}^+$  (pivot) such that:
- <u>α-property</u>: every positive edge that contains v is contained in K
- <u>β-property</u>: { $N \in \mathscr{C}^- \mid v \in N, N \subsetneq K$ }  $\cup$  {K} forms a total order w.r.t. inclusion with *K* as the smallest element

pivot for  $x_4$ 

 $Q() = A(x_1, x_2, x_3) \land U(x_3, x_4) \land \neg V(x_4) \land \neg R(x_2, x_3, x_4) \land \neg S(x_1, x_2, x_3, x_4)$ 

# The structure of signed-acyclicity

- A CQ is signed-acyclic **iff** it admits a signed-elimination sequence. At every step:
- find any signed-leaf *x* (with pivot *R*) 1.
- remove any relation with variables contained in R (<u> $\alpha$ -property</u>) 2.
- remove *x* from everywhere (<u>β-property</u>) 3.

 $Q() = A(x_1, x_2, x_3) \land U(x_3, x_4) \land \neg V(x_4) \land \neg R(x_2, x_3, x_4) \land \neg S(x_1, x_2, x_3, x_4)$ 

### A linear-time algorithm

We follow the signed-elimination sequence. At every step:

- find any signed-leaf *x* (with pivot *R*) 1.
- 2.  $(\alpha$ -property)
- 3. "Remove" x from every relation that contains it ( $\beta$ -property)

Item #3 is the challenging one!

Semi-join with *R* and remove any relation with variables contained in *R* 

### The Key Idea

- $Q() = A(x_1) \wedge B(x_2) \wedge \neg R(x_1, x_2)$ • We cannot afford to scan A for every value of B
  - a 3 4 5 e

start

• We build a data structure that encodes the "skips"



#### $Q() = A(x_1) \wedge B(x_2) \wedge \neg R(x_1, x_2)$





To "project out"  $x_1$  from *R*, we only keep the values that generate no answer (i.e.  $\{d\}$ )

b

d

 $Q'() = B(x_2)$ 

$$B = R \begin{bmatrix} b \\ b \\ c \end{bmatrix}$$

$$A \neg R(x_2) \qquad C \\ d \\ e \end{bmatrix}$$

 $Q(x_1, x_2) = A(x_1) \land B(x_2) \land \neg R(x_1, x_2)$ 





The skipping data structure can also be used to enumerate all results with constant delay

#### A linear-time characterization

For a signed-acyclic CQ with negation with input size *N*:

- if it is **Boolean**, it can be evaluated in linear time O(N)1.
- if it is **full**, the answers can be enumerated with constant delay after 2. linear-time preprocessing, with total time O(N + OUT)

Moreover, the algorithms have polynomial combined complexity

#### What about projections?

$$Q(\mathbf{x}_F) = \bigwedge_{K \in \mathscr{C}^+} K$$

If the signed hypergraph ([*n*],  $\mathscr{E}^+$ ,  $\mathscr{E}^- \cup \{F\}$ ) is signed-acyclic, the output can be enumerated with constant delay after linear-time preprocessing • This naturally captures the notion of **free-connex** CQs

- under widely believed conjectures

 $R_K(\mathbf{x}_K) \wedge \bigwedge \neg R_K(\mathbf{x}_K)$  $K \in \mathscr{E}^{-}$ 

• Any CQN not in this class does not admit a linear-time algorithm



# Counting: example

#### $#Q(x_2) = A(x_1) \land B(x_2) \land \neg R(x_1, x_2)$





For every value of B, count the number of nodes from A that are not connected with it

#### $#Q(x_2) = A(x_1) \land B(x_2) \land \neg R(x_1, x_2)$



start

- We can count by using the skipping DS to find the correct intervals and then compute the partial counts: a:[1-5]
- *b* : [3]



D

#### c: [1-2], [4-5]

#### $#Q(x_2) = A(x_1) \land B(x_2) \land \neg R(x_1, x_2)$





• We need to compute the partial sums

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• We can build a data structure in linear time such that we can calculate each partial sum in constant time (OFFLINE PARTIAL SUMS)

Idea: 
$$\sum_{i=u}^{v} x_i = \sum_{i=1}^{v} x_i - \sum_{i=1}^{v} x_i$$

 $#Q(x_2) = \bigoplus_{x_1} A(x_1) \wedge B(x_2) \wedge \neg R(x_1, x_2)$ 





- For any aggregation, where  $\oplus$  forms a semigroup • we can compute the partial sums in constant time • but we need preprocessing time  $O(N \cdot \alpha(N))$ •  $\alpha(N)$  is the inverse Ackermann function • uses deep results for RangeSum [Yao '82, Chazelle '91]

- Semiring structure  $(\mathbf{D}, \bigoplus, \bigotimes, \mathbf{0}, \mathbf{1})$
- the list has value **0**
- the list has the same default value  $c \neq \mathbf{0}$

# **Aggregation in Arbitrary CQNs**

 $Q(\mathbf{x}_F) = \bigoplus_{\mathbf{x}_{[n]\setminus F}} \bigotimes_{K \in \mathscr{E}^+} R_K(\mathbf{x}_K) \wedge \bigotimes_{K \in \mathscr{E}^-} R_K(\mathbf{x}_K)$ 

positive negative factor

factor

• <u>positive factor</u>: a list of tuples with their value in **D**; any tuple outside

• <u>negative factor</u>: a list of tuples with their value in **D**; any tuple outside

after preprocessing time  $O(N \cdot \alpha(N))$ 

- If the semiring has an additive inverse, the preprocessing time is O(N)
- The general algorithm follows the elimination sequence, but maintaining the aggregates becomes very complex

### Aggregation in Arbitrary CQNs

#### $Q(\mathbf{x}_F) = \bigoplus_{\mathbf{x}_{[n]\setminus F}} \bigotimes_{K \in \mathscr{E}^+} R_K(\mathbf{x}_K) \wedge \bigotimes_{K \in \mathscr{E}^-} R_K(\mathbf{x}_K)$

For any semiring, if the signed hypergraph ( $[n], \mathscr{E}^+, \mathscr{E}^- \cup \{F\}$ ) is freeconnex signed-acyclic, the output can be enumerated with constant delay

# Other Remarks

# Query Difference

with the same output schema:  $Q = Q_1 - Q_2$  [Hu & Wang '23]

$$Q = \left( R(x_1, x_2) \land S(x_2, x_3, x_4) \right) - \left( T(x_1, x_2) \land U(x_2, x_3) \right)$$
  
=  $\left( R(x_1, x_2) \land S(x_2, x_3, x_4) \land \neg T(x_1, x_2) \right) \cup \left( R(x_1, x_2) \land S(x_2, x_3, x_4) \land \neg U(x_2, x_3) \right)$ 

Since both resulting CQNs are signed-acyclic, we can enumerate their union with constant-delay enumeration after linear time preprocessing

Our techniques also characterize the linear-time behavior for the difference of two CQs

#### **Relational Division**

- Suppose we want to compute relational division: R(x, y)/S(x)
- We can rewrite using RA:  $\pi_y(R) \pi_y((\pi_y(R) \times S) R)$
- Define  $R'(y) = \pi_y(R)$ , which can be computed in linear time
- The RHS of the difference is the query  $Q(y) = R'(y) \wedge S(x) \wedge \neg R(x, y)$  which is freeconnex signed-acyclic and thus can be computed in linear time!

Corollary: the division operator can be computed in linear time

# **Open Questions**

- What are the appropriate measures of width to have tractability for CQNs?
- nest-set width [Lanzinger '21]
- generalizations of fractional hyper tree width?
- Do our algorithms translate to practice?
- query rewriting techniques [Hu & Wang '23]
- data structure implementation