Query Processing and Optimization

(Slides adapted from Thomas Eiter, Leonid Libkin and Werner Nutt)

Foundations of Databases 1

Query Processing and Optimization

- Query optimization: finding a good way to evaluate a query
- Queries are declarative, and can be translated into procedural languages in more than one way
- Hence one has to choose the best (or at least good) procedural query
- This happens in the context of query processing
- A query processor turns queries and updates into sequences of operations on the database

Query Processing and Optimization Stages

- Queries are translated into an extended relational algebra (operator + execution method): Which algebra expressions will allow for an efficient execution?
- Algebraic operators can be implemented by different methods (Examples?): Which algorithm should be used for each operator?
- How do operators pass data (write into main memory, write on disk, pipeline into other operators)?

Issues:

- Translate the query into extended relational algebra ("query plans")
- Esimate the execution of the plans: We need to know how data is stored, how it accessed, how large are intermediate results, etc.

Decisions are based on general guidelines and statistical information

Query Processing and Optimization

Foundations of Databases 3

Overview of Query Processing

Start with a declarative query:

```
SELECT R.A, S.B, T.E
FROM R,S,T
WHERE R.A>5 AND S.B<3 AND T.D=T.E
```

• Translate into an algebra expression:

$$\pi_{R.A,S.B,T.E}(\sigma_{R.A>5\land S.B<3\land T.D=T.E}(R\bowtie S\bowtie T))$$

• Optimization step: rewrite to an equivalent but more efficient expression:

$$\pi_{R,A,S,B,T,E}(\sigma_{A>5}(R) \bowtie \sigma_{B<3}(S) \bowtie \sigma_{D=E}(T)))$$

Why may this be more efficient? Is there a still more efficient plan?

Optimization by Algebraic Equivalences

ullet Given a relational algebra expression E, find another expression E' equivalent to E that is easier (faster) to evaluate.

- ullet Basic question: Given two relational algebra expressions E_1 , E_2 , are they equivalent?
- If there were a method to decide equivalence, we could turn it into one to decide whether an expression E is empty, i.e., whether $E(\mathbf{I})=\emptyset$ for every instance \mathbf{I} . How does one show this?
- ullet Problem: Testing whether $E\equiv\emptyset$ is undecidable for expressions of full relational algebra i.e., including union and difference.
- Good news:

We can still list some useful equalities.

Equivalence (and emptiness testing) is decidable for important classes of

Query Processing and Optimization

Foundations of Databases 5

queries (SPJR, SPC).

Optimization by Algebraic Equivalences (cntd)

Systematic way of query optimization: Apply equivalences

- ullet and imes are commutative and associative, hence applicable in any order
- \bullet Cascaded projections can be simplified: If the attributes A_1,\dots,A_n are among B_1,\dots,B_m , then

$$\pi_{A_1,...,A_n}(\pi_{B_1,...,B_m}(E)) = \pi_{A_1,...,A_n}(E)$$

• Cascaded selections might be merged:

$$\sigma_{C_1}(\sigma_{C_2}(E)) = \sigma_{C_1 \wedge C_2}(E)$$

 \bullet Commuting selection with join. If c only involves attributes from E_1 , then

$$\sigma_C(E_1 \bowtie E_2) = \sigma_C(E_1) \bowtie E_2$$

Query Processing and Optimization

Foundations of Databases

7

- Optimization by Algebraic Equivalences (contd)
- Rules combining σ, π with ∪ and \.
 Commuting selection and union:

$$\sigma_C(E_1 \cup E_2) = \sigma_C(E_1) \cup \sigma_C(E_2)$$

- Commuting selection and difference:

$$\sigma_C(E_1 \setminus E_2) = \sigma_C(E_1) \setminus \sigma_C(E_2)$$

- Commuting projection and union:

$$\pi_{A_1,\ldots,A_n}(E_1 \cup E_2) = \pi_{A_1,\ldots,A_n}(E_1) \cup \pi_{A_1,\ldots,A_n}(E_2)$$

– Question: what about projection and difference? Is $\pi_A(E_1 \setminus E_2)$ equal to $\pi_A(E_1) \setminus \pi_A(E_2)$?

• etc

Optimization of Conjunctive Queries

- Reminder:
 - Conjunctive queries
 - = SPJR queries
 - simple SELECT-FROM-WHERE SQL queries
 (only AND and (in)equality in the WHERE clause)
- Extremely common, and thus special optimization techniques have been developed
- Reminder: for two relational algebra expressions E_1, E_2 ,

"
$$E_1 = E_2$$
" is undecidable.

- But for conjunctive queries, even $E_1 \subseteq E_2$ is decidable.
- Main goal of optimizing conjunctive queries: reduce the number of joins.

Query Processing and Optimization

Foundations of Databases 9

Optimization of Conjunctive Queries: An Example

- ullet Given a relation R with two attributes A,B
- Two SQL queries:

Q1 Q2

SELECT DISTINCT R1.B, R1.A SELECT DISTINCT R3.A, R1.A FROM R R1, R R2 FROM R R1, R R2, R R3 WHERE R2.A=R1.B WHERE R1.B=R2.B AND R2.B=R3.A

- Are they equivalent?
- If they are, we can save one join operation.
- In relational algebra:

$$Q_1 = \pi_{2,1}(\sigma_{2=3}(R \times R))$$

$$Q_2 = \pi_{5,1}(\sigma_{2=4 \land 4=5}(R \times R \times R))$$

Optimization of Conjunctive Queries (contd)

- ullet We will show that Q_1 and Q_2 are equivalent!
- We cannot do this by using our equivalences for relational algebra expression.
 (Why?)
- Alternative idea: CQs are like patterns that have to be matched by a database. If a CQ returns an answer over a db, a part of the db must "look like" the query.
- Use rule based notation to find a representation of part of the db:

$$Q_1(x,y) := R(y,x), R(x,z)$$

 $Q_2(x,y) := R(y,x), R(w,x), R(x,u)$

Query Processing and Optimization

Foundations of Databases 11

Containment is a Key Property

Definition. (Query containment) A query Q is *contained* in a query Q' (written $Q\subseteq Q'$) if $Q(\mathbf{I})\subseteq Q'(\mathbf{I})$ for every database instance \mathbf{I} .

Translations:

- If we can decide containment for a class of queries, then we can also decide equivalence.
- If \mathcal{Q} is a class of queries that is closed under intersection, then the containment problem for \mathcal{Q} can be reduced to the equivalence problem for \mathcal{Q} .

Checking Containment of Conjunctive Queries: Ideas

- \bullet We want to check containment of two CQs $Q,\,Q'$
- We consider CQs without equalities and inequalities
- We also assume that Q and Q' have the same arity and identical vectors of head variables, that is, they are defined as $Q(\vec{x}) := B$, $Q'(\vec{x}) := B'$
- ullet Intuition: a conjunctive query $Q(\vec{x}):-B$ returns an answer over instance ${f I}$ if ${f I}$ "matches the pattern B"
- Instead of checking containment over all instances, we consider only finitely many test instances (or better, one!)

Query Processing and Optimization

Foundations of Databases

13

Tableau Notation of Conjunctive Queries

- Tableaux notation blurs the distinction between query and database instance
- We first consider queries over a single relation

$$Q_1(x,y) := R(y,x), R(x,z)$$

 $Q_2(x,y) := R(y,x), R(w,x), R(x,u)$

Tableaux:

_A	В		Α	В	
у	Х		у	х	
Х	z		w	х	
Х	у	← answer line	х	u	
			Х	У	← answer line

Variables in the answer line (or summary) are called "distinguished"

Tableau Homomorphisms

- Tableaux (as well as database instances or first order structures) are compared by homomorphisms
- \bullet Idea: T^\prime is more general than T if there is a homomorphism from T^\prime to T
- Reminder: Terms are variables or constants
- ullet A homomorphism δ from Tableau T_1 to tableau T_2 is a mapping

$$\delta \colon \{ \text{variables of } T_1 \} \to \{ \text{terms of } T_2 \}$$

such that

- $\delta(x) = x$ for every distinguished x
- if $(t_1,...,t_k)$ is a row in T_1 , then $(\delta(t_1),...,\delta(t_k))$ is a row in T_2 , where δ is extended to constants such that $\delta(a)=a$ for every constant a

Query Processing and Optimization

Foundations of Databases 15

Tableaux for Queries with Multiple Relations

- ullet So far we assumed that there is only one relation R, but what if there are many?
- Construct a tableau for the query:

$$Q(x,y)$$
:- $S(x,y), R(y,z), R(y,w), R(w,y)$

• We create rows for each relation:

 Formally, a tableau is just a database where variables can appear in tuples, plus a set of distinguished variables.

Tableaux Homomorphisms: General Case

- ullet Let T_1 , T_2 be tableaux with the same distinguished variables
- ullet A homomorphism δ from T_1 to T_2 is a mapping

$$\delta \colon \{ \text{variables of } T_1 \} \ \to \ \{ \text{terms of } T_2 \}$$

such that

- $\delta(a) = a$ for every constant
- $\delta(x)=x$ for every distinguished variable
- if \vec{t} is a row of R in T_1 , then $\delta(\vec{t})$ is a row of R in T_2 , for every relation R

Query Processing and Optimization

Foundations of Databases 17

The Homomorphism Theorem for Tableaux

Homomorphism Theorem: Let Q, Q' be two conjunctive queries without equalities and inequalities that have the same distinguished variables and let T, T' be their tableaux. Then

 $Q \subseteq Q' \, \Leftrightarrow \text{there exists a homomorphism from } T' \text{ to } T$

Proof: Necessity

- ullet Assume that $Q\subseteq Q'$
- ullet Note that T, ignoring the declaration of distinguished variables, can be seen as a database instance
- Note also that $\vec{x} \in Q(T)$
- $\bullet \; \operatorname{Since} \, Q \subseteq Q',$ it follows that $\vec{x} \in Q'(T)$
- ullet There is a valuation u such
 - ν satisfies Q' over T
 - $\nu(\vec{x}) = \vec{x}$
- \bullet Thus, defining $\delta(y)=\nu(y)$ for every variable y in Q' , we obtain a homomorphism from T' to T

Query Processing and Optimization

Foundations of Databases 19

Proof: Sufficiency

- \bullet Assume there is a homomorphism δ from T' to T. We will show that $Q\subseteq Q'$
- ullet Let ${f I}$ be a db instance. We show that $Q({f I})\subseteq Q'({f I})$
- ullet Let $ec{a}\in Q(\mathbf{I}).$ We show that $ec{a}\in Q'(\mathbf{I})$
- ullet There is a valuation $u\colon \{ ext{ variables of }Q\} o \mathbf{dom}(\mathbf{I}) ext{ such that}$
 - ν satisfies B, the body of Q (that is, " $\nu(B)\subseteq \mathbf{I}$ " in logic programming perspective)
 - $\nu(\vec{x}) = \vec{a}$

Proof: Sufficiency (cntd)

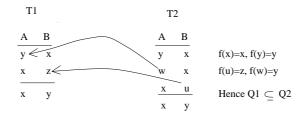
ullet Let $u':=
u\circ\delta.$ Then u' is a valuation for Q' with the following properties.

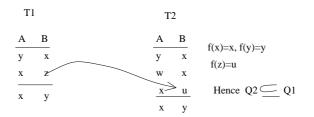
- First, $\nu'(\vec{x}) = \nu(\delta(\vec{x})) = \nu(\vec{x}) = \vec{a}$.
- ullet Moreover, if $R(\vec{t})$ is an atom of B', then
 - $\delta(R(\vec{t}))=R(\delta(\vec{t}))$ is an atom of B, since δ is a homomorphism, and
 - $\nu'(R(\vec{t}))=\nu(\delta(R(\vec{t})))=\nu(R(\delta(\vec{t})))$ is an atom in ${f I}$, since ν satisfies B
- ullet Hence, $ec{a} \in Q'(\mathbf{I})$ and the other claims are shown as well.

Query Processing and Optimization

Foundations of Databases 21

Applying the Homomorphism Theorem: $Q_1 \equiv Q_2$





Query Containment: Exercise

Find all containments and equivalences among the following conjunctive queries:

$$\begin{array}{llll} q_1(x,y) & \coloneq & r(x,y),\, r(y,z),\, r(z,w) \\ \\ q_2(x,y) & \coloneq & r(x,y),\, r(y,z),\, r(z,u),\, r(u,w) \\ \\ q_3(x,y) & \coloneq & r(x,y),\, r(z,u),\, r(v,w),\, r(x,z),\, r(y,u),\, r(u,w) \\ \\ q_4(x,y) & \coloneq & r(x,y),\, r(y,3),\, r(3,z),\, r(z,w) \end{array}$$

Query Processing and Optimization

Foundations of Databases 23

Query Containment: Complexity

- ullet Given two conjunctive queries, how hard is it to test whether $Q_1\subseteq Q_2$?
- It is easy to transform them into tableaux, from either SPJ relational algebra queries, or SQL queries, or rule-based queries.
- However, a polynomial algorithm for deciding "equivalence" is unlikely to exist:

Theorem. Given two tableaux, deciding the existence of a homomorphism between them is NP-complete.

 In practice, query expressions are small, and thus conjunctive query optimization is nonetheless feasible in polynomial time

Minimizing Conjunctive Queries

Goal: Given a conjunctive query Q, find an equivalent conjunctive query Q' with the minimum number of joins.

Questions: How many such queries can exist? How different are they?

Assumption: We consider only CQs without equalities and inequalities.

We call these queries simple conjunctive queries (SCQs).

If nothing else is said in this chapter, CQs are simple CQs

Terminology: If

$$Q(\vec{x}) := R_1(\vec{u}_1), \ldots, R_k(\vec{u}_k)$$

is a CQ, then Q^\prime is a **subquery** of Q if Q^\prime is of the form

$$Q'(\vec{x}) := R_{i_1}(\vec{u}_{i_1}), \dots, R_{i_l}(\vec{u}_{i_l})$$

where $1 \le i_1 < i_2 < \ldots < i_l \le k$.

Query Processing and Optimization

Foundations of Databases 25

Minimization: Background Theory

Proposition 1. Let q be a SCQ with n atoms and q' be an equivalent SCQ with m atoms where m < n. Then there exists a subquery q_0 of q such that q_0 has at most m atoms in the body and q_0 is equivalent to q.

Proposition 2. Let q and q' be two equivalent minimal SCQs. Then q and q' are identical up to renaming of variables.

Conclusions:

- There is essentially one minimal version of each SCQ Q.
- We can obtain it by dropping atoms from *Q*'s body.

An Algorithm for Minimizing SCQs

Given a conjunctive query Q, transform it into a tableau T.

Minimization algorithm:

```
T':=T; repeat until no change \text{choose a row } a \text{ in } T'; \text{if there is a homomorphism } \delta\colon T'\to T'\setminus\{a\} \text{then } T':=T'\setminus\{a\} end
```

Output: The query Q^\prime corresponding to the tableau T^\prime

Query Processing and Optimization

Foundations of Databases 27

Questions about the Algorithm

- Does it terminate?
- Is Q' equivalent to Q?
- ullet Is Q^\prime of minimal length among the queries equivalent to Q?

Minimizing SPJ/Conjunctive Queries: Example

- R with three attributes A, B, C
- SPJ query

$$Q = \pi_{AB}(\sigma_{B=4}(R)) \bowtie \pi_{BC}(\pi_{AB}(R) \bowtie \pi_{AC}(\sigma_{B=4}(R)))$$

• Translate into relational calculus (instead of normalizing):

$$(\exists z_1 \ R(x, y, z_1) \land y = 4) \land \exists x_1 \ ((\exists z_2 \ R(x_1, y, z_2)) \land (\exists y_1 \ R(x_1, y_1, z) \land y_1 = 4))$$

• Simplify, by substituting the constant, and putting quantifiers forward:

$$\exists x_1, z_1, z_2 (R(x, 4, z_1) \land R(x_1, 4, z_2) \land R(x_1, 4, z) \land y = 4)$$

• Conjunctive query:

$$Q(x, y, z) := R(x, 4, z_1), R(x_1, 4, z_2), R(x_1, 4, z), y = 4$$

Query Processing and Optimization

Foundations of Databases

29

• Tableau T:

$$egin{array}{ccccccc} A & B & C \\ \hline x & 4 & z_1 \\ x_1 & 4 & z_2 \\ \hline x & 4 & z \\ \hline \end{array}$$

Minimizing SPJ/Conjunctive Queries (contd)

ullet Minimization, step 1: Is there a homomorphism from T to

• Answer: No. For any homomorphism δ , we have $\delta(x)=x$ (why?), thus the image of the first row is not in the smaller tableau.

Minimizing SPJ/Conjunctive Queries (contd)

- ullet Answer: Yes. Homomorphism δ : $\delta(z_2)=z$, all other variables stay the same.
- The new tableau is not equivalent to

ullet Because $\delta(x)=x, \delta(z)=z,$ and the image of one of the rows is not present.

Minimizing SPJ/Conjunctive Queries (contd)

Query Processing and Optimization

Foundations of Databases

31

$$egin{array}{c|cccc} A & B & C \\ \hline x & 4 & z_1 \\ \hline x_1 & 4 & z \\ \hline \end{array}$$

• Back to conjunctive query:

$$Q'(x, y, z) := R(x, y, z_1), R(x_1, y, z), y = 4$$

• An SPJ query:

$$\sigma_{B=4}(\pi_{AB}(R) \bowtie \pi_{BC}(R))$$

• Pushing selections:

$$\pi_{AB}(\sigma_{B=4}(R)) \bowtie \pi_{BC}(\sigma_{B=4}(R))$$

Review of the Journey

We started with

$$\pi_{AB}(\sigma_{B=4}(R)) \bowtie \pi_{BC}(\pi_{AB}(R) \bowtie \pi_{AC}(\sigma_{B=4}(R)))$$

- Translated into a conjunctive query
- Built a tableau and minimized it
- Translated back into conjunctive query and SPJ query
- Applied algebraic equivalences and obtained

$$\pi_{AB}(\sigma_{B=4}(R)) \bowtie \pi_{BC}(\sigma_{B=4}(R))$$

• Savings: one join.

Query Processing and Optimization

Foundations of Databases 33

Minimization of Conjunctive Queries: Multiple Relations

• We consider again the query:

$$Q(x,y)$$
:- $B(x,y), R(y,z), R(y,w), R(w,y)$

• The tableau was:

Minimization with Multiple Relations

ullet The algorithm is the same as before, but one has to try rows in different relations. Consider the homomorphism where $\delta(z)=w$, and δ is the identity for all other variables. Applying this to the tableau for Q yields

$$B: \begin{array}{c|cccc} A & B \\ \hline x & y \end{array} \qquad \qquad R: \begin{array}{c|cccc} A & B \\ \hline y & w \\ \hline w & y \end{array}$$

- This can't be further reduced, as for any homomorphism δ , $\delta(x) = x$, $\delta(y) = y$.
- ullet Thus Q is equivalent to

$$Q'(x,y) := B(x,y), R(y,w), R(w,y)$$

• One join is eliminated.

Query Processing and Optimization

Foundations of Databases 35

Query Optimization and Functional Dependencies

- · Additional equivalences can be inferred if integrity constraints are known
- We consider here functional dependencies
- Example: Let R have attributes A, B, C. Assume that R satisfies $A \to B$.
- Then it holds that

$$R = \pi_{AB}(R) \bowtie \pi_{AC}(R)$$

- Tableaux can help with these optimizations!
- ullet $\pi_{AB}(R)\bowtie\pi_{AC}(R)$ as a conjunctive query:

$$Q(x, y, z)$$
:- $R(x, y, z_1), R(x, y_1, z)$

Query Optimization and Functional Dependencies (contd)

$$egin{array}{ccccc} {\sf A} & {\sf B} & {\sf C} \\ \hline x & y & z_1 \\ \hline x & y_1 & z \\ \hline x & y & z \end{array}$$

• Using the FD $A \rightarrow B$ infer $y = y_1$

- ullet And this says that the query is equivalent to Q'(x,y,z):-R(x,y,z), that is, R
- This is known as the "chase" technique

Query Processing and Optimization

Foundations of Databases 37

Query Optimization and Functional Dependencies (contd)

- General idea: simplify the tableau using functional dependencies and then minimize.
- $\bullet\,$ Given: a conjunctive query Q , and a set of FDs F
- Algorithm:
 - Step 1. Construct the tableau T for Q
 - Step 2. Apply algorithm CHASE(T, F)
 - Step 3. Minimize output of $\mathsf{CHASE}(T,F)$
 - Step 4. Construct a query from the tableau produced in Step 3

The CHASE

We assume that all FDs are of the form $X \to A$, where A is a single attribute. For simplicity, we also assume that the tableau has only a single relation. The generalisation is straightforward.

```
for each X \to A in F do for each t_1, t_2 in T such that t_1[X] = t_2[X] and t_1[A] \neq t_2[A] do case t_1[A], t_2[A] of one nondistinguished variable \Rightarrow replace the nondistinguished variable by the other term one distinguished variable, one distinguished variable or constant \Rightarrow replace the distinguished variable by the other term two constants \Rightarrow output \bot and STOP end
```

Query Processing and Optimization

Foundations of Databases 39

Query Optimization and Functional Dependencies: Example 2

- $\bullet \ R \text{ is over } A,B,C; \ \ F=\{\, B\to A\,\}$
- $Q = \pi_{BC}(\sigma_{A=4}(R)) \bowtie \pi_{AB}(R)$
- Q as a conjunctive query:

$$Q(x, y, z) := R(4, y, z), R(x, y, z_1)$$

• Tableau:

• Final result: Q(x, y, z):- R(x, y, z), x = 4, that is, $\sigma_{A=4}(R)$.

Query Optimization and Functional Dependencies: Example 3

• Same R and F; the query is:

$$Q = \pi_{BC}(\sigma_{A=4}(R)) \bowtie \pi_{AB}(\sigma_{A=5}(R))$$

• As a conjunctive query:

$$Q(x, y, z) := R(4, y, z), R(x, y, z_1), x = 5$$

• Tableau:

- Final result: ⊥ (the empty query)
- ullet This equivalence does not hold *without* the FD B
 ightarrow A

Query Processing and Optimization

Foundations of Databases 41

Query Optimization and Functional Dependencies: Example 4

- Sometimes simplifications are quite dramatic
- Same R, FD is $A \rightarrow B$, the query is

$$Q = \pi_{AB}(R) \bowtie \pi_A(\sigma_{B=4}(R)) \bowtie \pi_{AB}(\pi_{AC}(R) \bowtie \pi_{BC}(R))$$

• Convert into conjunctive query:

$$Q(x,y) := R(x,y,z_1), R(x,y_1,z), R(x_1,y,z), R(x,4,z_2)$$

Query Optimization and Functional Dependencies: Example 4 (contd)

• Tableau:

Query Processing and Optimization

Foundations of Databases 43

Query Optimization and Functional Dependencies: Example 4 (contd)

- ullet or, equivalently $\pi_{AB}(\sigma_{B=4}(R)).$
- Thus,

$$\pi_{AB}(R) \bowtie \pi_A(\sigma_{B=4}(R)) \bowtie \pi_{AB}(\pi_{AC}(R) \bowtie \pi_{BC}(R)) = \pi_{AB}(\sigma_{B=4}(R))$$
 in the presence of FD $A \to B$.

- Savings: 3 joins!
- This cannot be derived by algebraic manipulations, nor conjunctive query minimization without using CHASE.

Questions about the CHASE

- Does the CHASE algorithm terminate? What is the run time?
- What is the relation between a tableau and its CHASE'd version?
- Query containment wrt a set of FD's:
 - How can we define this problem?
 - Can we decide this problem?
- Query minimsation wrt to a set of FDs
- Consider SCQs: we know from previous exercises that all such queries are satisfiable. Is the same true if we assume only database instances that satisfy a given set of FDs F?

Query Processing and Optimization

Foundations of Databases 45

Conjunctive Queries with Equalities and Inequalities

- Equality / Inequality atoms x = y, x = a, $x \neq z$, etc
- \bullet Let T,T' be the tableaux of the parts of conjunctive queries Q and Q' with ordinary relations
- Sufficiency: $Q\subseteq Q'$ if there exists a homomorphism $\delta\colon T'\to T$ such that for each (in)equality atom $t_1\theta t_2$ in Q', we have that $\delta(t_1)\theta\delta(t_2)$ is logically implied by the equality and inequality atoms in Q
- However, existence of a homomorphims is no more a necessary condition for containment.
- It holds under certain conditions, though.

Note: Deciding whether a set of equality / inequality atoms ${\cal A}$ logically implies an equality / inequality atom is easy.

Query Homomorphism: Definition

Instead of tableau homomorphims, one often defines query homomorphisms.

An **homomorphism** from

$$q'(\vec{x}) := R', C'$$

to

$$q(\vec{x}) := R, C$$

is a substitution δ such that

- $\delta(\vec{x}) = \vec{x}$
- $\delta(R') \subseteq R$
- $C \models \delta(C')$.

Here, $\delta(\vec{x})$, $\delta(R')$ and $\delta(C')$ are the extensions of δ to complex syntactic entities.

Note that we view $R,\,R',\,C$, and C' as sets of atoms.

Query Processing and Optimization

Foundations of Databases 47

Query Homomorphism: Example

$$Q'(x)$$
 :- $P(x,y), R(y,z),$
 $y \le 3$
 $Q(x)$:- $P(x,w), P(x,x), R(x,u),$
 $w > 5, x < 2$

An homomorphism from Q^\prime to Q is

$$\delta(x) = x, \ \delta(y) = x, \ \delta(z) = u.$$

Homomorphisms: The General Case

Existence of a homomorphism is not a necessary, but a sufficient condition for containment.

Theorem: Let $Q,\,Q'$ be two conjunctive queries, possibly with comparisons and inequalities, such that Q and Q' have the same distinguished variables. Then

 $Q\subseteq Q'$ if there exists a homomorphism from Q' to Q.

Query Processing and Optimization

Foundations of Databases 49

Containment of Queries With Comparisons

Classical example:

$$Q'()$$
 :- $P(u, v), u \le v$
 $Q()$:- $P(y, z), P(z, y)$

- Q is contained in Q',
- ullet but there is no homomorphism from Q^\prime to Q.

Checking Containment of Queries with Comparisons: Idea

$$Q'()$$
 :- $P(u, v), u \le v$
 $Q()$:- $P(y, z), P(z, y)$

Replace Q with its linear expansion $(Q_L)_L!$

$$\begin{aligned} &Q_{\{\,y < z\,\}} & := & P(y,z), \, P(z,y), \, y < z \\ &Q_{\{\,y = z\,\}} & := & P(y,z), \, P(z,y), \, y = z \\ &Q_{\{\,y > z\,\}} & := & P(y,z), \, P(z,y), \, y > z \end{aligned}$$

(case analysis)

Query Processing and Optimization

Foundations of Databases 51

Klug's Theorem

Theorem (Klug 88): If

- -Q,Q' are conjunctive queries with comparisons
- $-(Q_L)_L$ is the **linear expansion** of Q,

then:

$$Q\subseteq Q'\quad\Leftrightarrow\quad\text{for every }Q_L\text{ in }(Q_L)_L,$$
 there is an homomorphism from Q' to Q_L

(analogous for disjunctive queries)

Containment with comparisons is $\Pi_2^P\text{-complete.}$

(van der Meyden)

Readings

- S. Abiteboul, R. Hull, and V. Vianu. Foundations of Databases. Addison-Wesley, 1995.
 - Chapter 6 (Sections 6.1 and 6.2 for containment, equivalence and minimization of conjunctive queries) and Chapter 8 (Section 8.4 for the case with functional dependencies).
- If you want to study the same problems for the case of queries with inequalities,
 see the papers:
 - Anthony C. Klug. On conjunctive queries containing inequalities. Journal of the ACM 35(1): 146-160 (1988)
 - Ron van der Meyden. The Complexity of Querying Indefinite Data about Linearly Ordered Domains. Journal of Computer and System Sciences. 54(1): 113-135 (1997)

Query Processing and Optimization