

Query Processings by Relational Algebra extended with Unification

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1 Introduction

Conventional deductive databases concentrate on fast retrieval of many fact-clauses through a few rule-clauses. Practical applications, however, will need a mass of rule-clauses and fast retrieval mechanisms for them. For this purpose, we propose *Relational Algebra extended with Unification* (RAU); a variant of relational algebra for handling unification as in [1].

This paper shows a large database of rule clauses, RAU-operators and their commutative laws.

2 Large rule-database

Deductive databases in this paper allow functor symbols. A rule-clause is abbreviated as a *rule* and a fact-clause as a *fact*. In general, deductive databases have two databases; a database of facts (factDB) and the other one of rules (ruleDB). We use two meta-predicates ruledb and demo in [2] for managing a ruleDB. ruledb(KB,Head,Body) says "a knowledgebase KB knows a rule Head:-Body". demo(T,Goal) says "a theory T proves Goal". In this paper, we say that a head-predicate p of a given rule p(..):-... is the rule's *kind*, and the rule *belongs to* or *expresses* its kind.

In practical applications, two cases enlarge a ruleDB; either there are many *kinds* of rules, or many rules belong to one *kind*.

The latter case expresses "many different implementations for a common interface". In an object oriented paradigm, it is the case that a superclass *C* requires a common interface *p* and allows each subclass of *C* to implement *p* independently. Then if there are 10^4 subclasses, 10^4 rules belong to a kind *p*.

Figure 1 is an example of the latter case. store(Media, User, Data) is a common inter-

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% store(Media,User,Data):- Cond.
104 {
rules { store(type1(T), usa(P,cal(X)),
         image(noaa,A) ):- q1(T,P,X,A).
        store(type2(sub2(T)), usa(ic,C),
        : text(fl,A):- q2(T,C,A).
        :
% key(Data,Keyword,User):- Cond.
104 {
rules { key( text(S,A), K, usa(ic,X))
        : :- p1(S,A,K,X).
        key( image(S,A), story(K),
        : japan(ple,tokyo(X)) )
        : :- p2(S,A,K,X).
% query
q(X,Y):-
  ruledb(kb1,
    key(D, story(aaa), japan(P,X) ),
    Cond1),
  ruledb(kb2,
    store(M, usa(P,Y), D), Cond2 ),
  demo( to, (Cond1,Cond2)).

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Figure 1: an example of rule database

face of a class Media. It says "a User stores a Data in a Media". In Figure 1, subclasses and values of attributes in Media, User... are expressed by compound terms. e.g. User has a structure of *nation(group, city(idnumber))*. Rules expressing store are different from each other, depending on a type (i.e. subclass) of Media (e.g. visual terminal,...), properties of User (nation, group,..), and a type of Data (image, text, format,..). If there are 10^2 types of Media and 10^2 properties of User, 10^4 rules belong to the kind store. In the same way, key(Data, Keyword, User) is a common interface of a class Data. It says "a Data is registered as a Keyword by a User". Implementations of rules expressing key depend on a type of Data and properties of a User.

Most of queries are issued via only those common interfaces regardless of different implementations. The query in Figure 1 is written in meta-predicates. It retrieves applicable combinations of rules expressing "store and key" at first, and executes them.

R, T : meta-relations. A, B, C : attributeID.

$$\begin{array}{c}
 \frac{R[A \quad B]}{f(a,X) \quad p(X)} \quad \frac{T[A \quad C]}{X \quad q(X)} \\
 \swarrow \quad \quad \quad \swarrow \\
 r(f(a,X)):- p(X) \quad \quad \quad t(X):- q(X)
 \end{array}$$

- $R \bowtie T = \frac{[A \quad B \quad C]}{f(a,X) \quad p(X) \quad q(f(a,X))}$
- $\sigma_{A=g(X)} T = \frac{[A \quad C]}{g(X) \quad q(g(X))}$
- $I_B R = \frac{[A \quad B]}{f(a,b) \quad p(b)}$ where $p(b)$ is true.
- $\rho_{[f(A,B),C]} R = \frac{[A \quad B \quad C]}{a \quad X \quad p(X)}$

Figure 2: examples of RAU-operators

3 RAU-operators

A DBMS for both a large ruleDB and a large factDB must execute the query in Figure 1 as fast as possible. The query clarifies two requirements. One is a fast retrieval mechanism for a large ruleDB in case that many rules belong to one kind. The other is to avoid random accesses to a ruleDB in a disk when executing rules; because a large ruleDB may be stored in a disk.

A limited solution of the latter is a partial compilation; transformation of each rule to simpler ones which operate a factDB directly [3]. By this method, much more rules belong to one kind. e.g. a rule $p:-q, r.$ is compiled into 100 rules if q and r are compiled into 10 rules. Therefore a fast retrieval mechanism is fundamental for a large ruleDB.

Our approach is simple; At first, we compile in advance each rule into programs of a variant of relational algebra such as ERA in [4]. The variant must be able to deal with functor symbols. Then, in Figure 2, let $R[A,B]$ (or $T[A,C]$) be a set of rules expressing a common kind $r(A):- B.$ (or $t(A):- C.$). Set-operators are defined on these sets. We call this set of rules a *meta-relation* and these operators *Relational Algebra extended with Unification* (RAU). Their formal definitions are presented in [5,7]. In Figure 2, $I_B R[A,B]$ is a set of facts satisfying each rule in $R.$ $R[A,B] \bowtie T[A,C]$

is a set of rules expressing "r and t". RAU is also used for compiled expressions of rules. Queries are described as tree-forms of these operators, optimized, and executed by fast set-operation algorithms.

Because I-operator executes a set of modified relational algebra programs, it needs a global query optimization for common subexpression sharings such as in [6].

4 Commutative laws

Commutative laws of RAU-operators are forms of $exp1 =_w exp2.$ These laws are used for optimizing query trees.

Definition Assume that meta-relations R, T and tuples t, s are given. Then,

- $R \subset_w T \stackrel{\text{def}}{=} \forall t \in R, \exists s \in T, \exists \theta: \text{substitution, } t = s\theta.$
- $R =_w T \stackrel{\text{def}}{=} R \subset_w T \text{ and } T \subset_w R. \quad \square$

Commutative laws hold as follows [7]; (M, N, R are meta-relations. $1, 2, \dots$ are attributeID).

1. $\sigma_{p1 \wedge p2}(M \times N) =_w \sigma_{p1 \wedge p2}(\sigma_{p1} M \times \sigma_{p2} N),$ where $p1$ (or $p2$) is a selection-predicate including only attributes in M (or N).
2. $\sigma I M =_w I \sigma M.$
3. $\pi_1 I_3 M[1, 2, 3] =_w \pi_1 I_3 \pi_{1,3} M[1, 2, 3].$
4. $I_{a \wedge b}(M \bowtie N) =_w I_b((I_a M) \bowtie N) =_w I_a M \bowtie I_b N,$ where a (or b) is an attribute of M (or N).
5. $\rho_{tl} \pi_1 I_2 M[1, 2] =_w \pi_{vl} I_2 \rho_{[tl, 2]} M[1, 2],$ where tl is a list of compound term. vl is a list of distinct variable symbols in $tl.$

- [References] [1]. Morita, VLDB86, pp.52-59. [2]. Bowen, *amalgamating languages and meta language in logic programming*, Syracuse Univ, TR June, '81. [3]. Miyazaki, ICOT-TR183.'87. [4]. Zaniolo, VLDB85. pp.458-469. [5]. Ohmori, to be appeared in IWDM87. [6]. Sellis, SIGMOD86, pp.191-205. [7]. Ohmori, master thesis, Univ. of Tokyo. '87.