TRANSLATION AND OPTIMIZATION OF LOGIC QUERIES:
THE ALGEBRAIC APPROACH

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ABSTRACT

This paper presents an algebraic approach to translation and optimization of logic queries. We first develop a syntax directed translation from rules of function-free logic programs to algebraic equations; then we show solution methods for independent equations and for systems of interdependent equations. Such solutions define the operational and fixpoint semantics of function-free logic programs and queries. We also present algebraic optimization methods for "top-down" and "bottom-up" strategies; the former are useful if no initial binding is provided with the query, while the latter are useful if some arguments of the query are bound to constant values.

1. INTRODUCTION

In recent times, the combination of relational databases and logic programming (LP) has become a popular argument of research. The application of LP as query language of a relational database entails a relevant enrichment of the expressive power of traditional query languages; hence the database community looks at LP as a promising approach for posing complex (e.g. recursive or deductive) queries. At the same time, databases provide the technology for processing large collections of data in an efficient way, hence solving many of the problems posed by LP applications when they manage large amounts of information.

The efficient implementation of logic queries has been discussed in many recent papers, including Bancilhon86, Chandra82, Ceri86, Henschen84, Marque-Pucheu84, Sacca86, Ullman85. Major research directions are:

a. Designing "pure" LP languages, e.g. languages which do not incorporate procedural features, such as the dependency of the computation from the order of clauses and the use of special predicates. This trend has marked the adoption of "pure" Horn clauses as "standard" LP language, and consequently a certain resistance to Prolog.

b. Determining how function-free Horn Clauses can be efficiently executed. New formal models, such as "rule-goal" graphs, describe binding propagation among clauses. Several rules for "capturing" nodes of graphs have been defined; capturing a node is equivalent to evaluating the information associated to the corresponding clause or rule, deduced from the database.

c. Determining how LP programs can be made more efficient, by operating transformations from LP to LP. Techniques such as the "magic set", the "counting" and the "Eager" methods have been developed to this purpose.

We have focused our attention on the use of relational algebra at work on the same problems. While the idea of using relational algebra for executing logic queries is not new (see, among others, [Aho 79] and [Marque-Pucheu 84]), the major contribution of this paper is to give a systematic overview of how traditional algebra, extended by a closure operator, can be applied to solve logic queries.

2. MODELS FOR LOGIC PROGRAMMING AND RELATIONAL DATABASES

In this section, we give our definition and interpretation of function-free logic programs and queries; then, we introduce positive algebra extended with the closure operator.

2.1 function-free logic programs

A function-free logic program (FFLP) is a set of definite, function-free Horn clauses (i.e. clauses that contain exactly one positive literal); we use a Prolog-like syntax for clauses, which have their positive literal on the LHS and zero or more (negative) literals on the RHS, in conjunctive form. For instance, the following is a FFLP:

\[ \text{S(a,b).} \]
\[ \text{R(c,b).} \]
\[ \text{P(X):= P(Y), R(X,Y).} \]
\[ \text{Q(X,Y):= P(X), R(X,Z), S(Y,Z).} \]

We can better interpret a FFLP within the framework of databases if we consider terms appearing in the LHS of clauses as either database relations or computed relations. Ground clauses of the former are stored in an extensional database EDB; ground clauses of the latter are evaluated by executing the FFLP; thus, each computed relation appears as LHS of one or more clause. In the above example, \( R \) and \( S \) are database relations, \( P \) and \( Q \) are computed relations; the first two clauses are equivalent to assigning an instance to relations \( R \) and \( S \), and in general will not be present in FFLPs but will be stored in the EDB.

We generalize the class of database relations to include all those terms for which we have a function available for evaluating arithmetic predicates. For instance, the relation \( \text{PLUS}(X,Y,Z) \)
Each variable $X$ of a clause of a FFLP program is associated to a finite range $Rg(X)$. Ranges of variables of database relations are limited to the values existing in the instance of that relation or to the domain and co-domain of functions; all variables of computed relations are limited to a unique finite universe $U$ which includes all individuals possibly occurring in the DB.

An input goal to a FFLP is a clause consisting of a single literal, for instance:

$$\neg \varphi(a,X).$$

The solution of an input goal $P(t_1,...,t_n)$ is the set of all ground instances of $P(t_1,...,t_n)$, which are logical consequences of the FFLP; for example:

$$\text{Sol}(\text{FFLP}, \varphi(a,c)) = \{ \varphi(a,c) \mid (c \in U) \land (\text{FFLP} \cup \text{EDB} \Rightarrow \varphi(a,c)) \}$$

If the input goal is a literal with all places bound to constant values, then the solution space is reduced to the answers "yes" or "no";

$$\text{Sol}(\text{FFLP}, \varphi(c)) = \{ \text{"yes" if } (c \in U) \land (\text{FFLP} \cup \text{EDB} \Rightarrow \varphi(c)), \text{ "no" otherwise.} \}$$

2.2. Positive algebra with closure

We assume that the reader is familiar with the relational model and algebra as from [ULLMAN82].

**Positive relational algebra** (RA+) includes as primitive operators selection ($\sigma$), projection ($\pi$), cartesian product ($\times$), join ($\Join$), semi-join ($\bowtie$), and union ($\cup$); noticeably, it does not include the difference operator. We extend projection to include $\pi a.E$ (projection on the empty set) as an operator which can be applied to an algebraic expression $E$; $\pi a.E$ returns "yes" if $E \neq \emptyset$, "no" if $E = \emptyset$. $\emptyset$ denotes the empty relation of suitable degree.

The closure operator ($\text{O}^+$) is applied to an expression $E(X)$ of a (variable) relation $X$, provided that the schema of the result of $E(X)$ is the same as the schema of $X$. The language obtained by extending RA+ with the closure operator is denoted as ERA+. The operational definition of the closure operator is given by the following program $A_0$ which computes $O^+E(X)$:

$$A_0(E(X)):
\begin{align*}
& \text{OS} \leftarrow \Phi \\
& \text{REPEAT} \\
& \quad Y \leftarrow E(S) \\
& \quad S \leftarrow S \cup Y \\
& \quad \text{UNTIL } Y= S \\
& \text{RETURN } S
\end{align*}$$

The fixpoint semantics of the closure operator is also well defined. Let $P$ denote the set of all possible relations having the same domain as $X$:

$$P = \text{powerset}(\text{Dom}(X))$$

Then, $P$ is a complete lattice under the union and intersection operations, having as minimal element the empty relation and as maximal element $\text{Dom}(X)$.

Expressions of RA+ are monotone: (i.e. $X \subseteq Y \Rightarrow E(X) \subseteq E(Y)$), hence algorithm $A_0$ always terminates after a finite number of iterations, and $O^+E(X)$ is the minimal solution (i.e. the least fixpoint) of the equation $X=E(X)$:

$$O^+E(X) = \min \{ X \in P \mid X = E(X) \}$$

3. SYNTAX DIRECTED TRANSLATION FROM FUNCTION-FREE LOGIC PROGRAMS TO EXTENDED RELATIONAL ALGEBRA

A syntax-directed translation algorithm maps each clause of a FFLP into a correspondent disequation of RA+. Disequations are subsequently interpreted as equations under the Closed World Assumption (CWA) and solved using the closure operation.

3.1 Translation of individual clauses

Let a generic Horn clause of a FFLP have the following structure:

$$R : P(x_1,...,x_n) \leftarrow Q_1(\beta_1,...,\beta_k), ..., Q_h(\beta_s,...,\beta_m)$$

Then, the translation associates to $R$ an algebraic disequation:

$$\text{Expr}(Q_1,...,Q_h) \subset P$$

where $P$ is a computed relation correspondent to predicate $P$, and similarly $Q_i$ are either computed or database relations correspondent to predicates $Q_i$. The specification of the algorithm for the syntax directed translation requires defining some useful notation and two rewriting functions.

- $\text{occurs}(x,RHS)$, of sort boolean, is "true" if the term $x$ belongs to the RHS, "false" otherwise.
- $\text{const}(i)$ denotes the function returning, for each $x_i$ of the LHS, the index $j$ of the leftmost variable $\beta_j$ of the RHS such that $\beta_j=x_i$ if such a variable exists.
- $\text{var}(x_i)$ of sort boolean, is true if $x_i$ is a constant, false otherwise; $\text{const}(\beta_j)$ is similarly defined.
- $\text{newvar}(x)$, of sort boolean, is defined as "not $\text{const}(x)$".

Let $E$ denote a string, $x$ and $y$ denote symbols. Then $E\langle x,y \rangle$ denotes a new string in which the first occurrence of $x$ is substituted by $y$.

The syntax directed translation of $R$ into an algebraic disequation of RA+ is defined through two recursive rules which apply to the LHS and RHS of $R$ respectively.

The first rule deals with three special cases:

- bindings to constant values in the LHS;
- multiple occurrences of the same variable in the LHS;
- existence of a variable of the LHS that does not occur in the RHS.
Each such special case is reconducted, by suitable string transformations, to an equivalent case in which all positions of the LHS are bound to distinct variables. Then, the second rule is applied to the RHS; it simply generates all selection conditions due to constant bindings or replicated variables within the RHS, and then builds a cartesian product with all database or computed relations corresponding to terms of the RHS.

After applying the second rule, equivalence-preserving transformations can be applied to the resulting expression in RA+ to transform cartesian products into joins and to propagate selections to their operands, in the conventional way (see Ullman82). The notation $X_1=1..k Q_1$ indicates the cartesian product of relations $Q_1,..,Q_k$.

**rule 1.**

$$T(R) = \text{if } \exists i : \text{const}(x_i) \neq */\star/$$

then newvar($x_i$)

$$\text{return } T(LHS<\alpha_{1\star}x_i>) :- RHS, EQ(x_i, \alpha_{1\star})$$

elsif $\exists i,j : \alpha_{1\star}x_i \neq \alpha_{1\star}x_j, i < j$ /*/\star/

then newvar($x_i$)

$$\text{return } T(LHS<\alpha_{1\star}x_i>) :- RHS, EQ(x_i, \alpha_{1\star})$$

elsif $\exists i,j : \text{var}(x_i) \land \text{not occurs}(x_i, RHS) \neq */\star/$

then return $T'(\text{corr}(1),..,\text{corr}(n)) T'(RHS)$

end if;

**rule 2.**

$$T'(R) = \text{if } \exists i : \text{const}(\beta_i) \neq */\star/$$

then newvar($x_i$)

$$\text{return } \text{corr}(1),..,\text{corr}(n) T'(RHS)$$

elsif $\exists i,j : \beta_i = \beta_j, i < j$ /*/\star/

then newvar($x_i$)

$$\text{return } \text{corr}(1),..,\text{corr}(n) T'(RHS)$$

elsif $1 : \text{var}(\alpha) \land \text{not occurs}(\alpha, RHS) \neq */\star/$

then return $T'(\text{corr}(1),..,\text{corr}(n)) T'(RHS)$

end if;

Example. The following is a systematic application of the translation algorithm.

$P(X_1,Y_1,Z_1):-S(X_1,Y_1,R(Y_1,Z_1),EQ(N_1,X_1))$

1. by $T$, case (b)

$T(P(N_1,X_1,Z_1):-S(X_1,Y_1,R(Y_1,Z_1),EQ(N_1,X_1)))$

2. by $T$, last recursive call

$T_3(X_1,Y_1,Z_1,R(Y_1,Z_1),EQ(N_1,X_1))$

3. by $T$, cases (a) and (b)

$T_2(X_1,Y_1,Z_1,R(Y_1,Z_1),EQ(N_1,X_1))$

4. by $T$, last recursive call

$T_1(X_1,Y_1,Z_1,R(Y_1,Z_1),EQ(N_1,X_1))$

5. After pushing selection and join conditions

$T_1(X_1,Y_1,Z_1,R(Y_1,Z_1),EQ(N_1,X_1))$

6. final disjunction

$T_0(X_1,Y_1,Z_1,R(Y_1,Z_1),EQ(N_1,X_1)) \in P$

3.2. Transformations from disjunctions to equations in RA+

By effect of the translation rules explained above, we can turn each FLP into a set of algebraic disequations. For instance, the last 3 clauses of the FLP in Section 2.1 generate:

$$T_2(P \equiv_{1 \star} R) \in P$$

$$T_2 S \in P$$

$$T_{1,4}(P \equiv_{1 \star} R) \equiv_{3 \star} S \in Q$$

The Closed World Assumption (CWA, [Reiter78]) enables us to turn these disequations into equations. By the CWA, all facts which cannot be deduced by the application of the FLP to the database are false; hence, no fact about a generic term $P$ can be proved other than through existing disequations; hence the union of all RHS of disequations gives the algebraic equation required to compute $P$. The above example generates the system of equations:

$$P = T_1(P \equiv_{1 \star} R) U (T_2 S)$$

$$Q = T_{1,4}(P \equiv_{1 \star} R) \equiv_{3 \star} S$$

Marque-Pucheu et al. [Marque_Pucheu84] show a transformation from logic programs into equations which does not directly use relational algebra.

3.3. Solution of independent equations

Solving a system of equations is easier if each equation is independent from the others (or can be made independent by suitable substitutions); an equation is independent when it does not contain computed relations in the RHS other than the one in the LHS. For the solution of independent equations, two cases are given:

a. The RHS contains only database relations; in this case, the solution of the equation is simply given by the evaluation of the expression in RA+. This happens when rules for the computed relation in the FLP are nonrecursive.

b. The RHS contains one or more occurrences of the computed relation $CR$; in this case, the definition of the closure operator as fixpoint of algebraic equations enables us to build the solution as follows:

$$sol(CR) = 0 \land CR \land \text{RHS}(CR)$$

In the above example (Sect. 3.2), the first equation is independent and can be solved as:

$$sol(P) = P \land (T_1(P \equiv_{1 \star} R) U (T_2 S))$$

The second equation for $Q$ depends on $P$, however we can suspend its evaluation until we have solved the equation for $P$. Then we consider $P$ as a fixed (i.e. database) relation, and the second equation falls in case (a) above:

$$sol(Q) = T_{1,4}(sol(P) \equiv_{1 \star} R) \equiv_{3 \star} S$$

We can now interpret input goals as suitable expressions on the algebraic solutions; for instance:

$$?-P(X). \iff sol(P)$$

$$?-P(a). \iff T_{1,4}(sol(P) \equiv_{1 \star} a) sol(P)$$

$$?-Q(a,X). \iff G_{1,a} sol(Q)$$

$$?-Q(a,b). \iff T_{1,4}(sol(P) \equiv_{1 \star} a \land z=b sol(Q)$$
3.4. Reduction by substitutions

We have seen that by suitable substitutions it is sometimes possible to reduce a system of mutually interdependent equations to a system of equations which are independently resolvable, when a certain order of evaluation is observed. We call these systems of equations reducible by substitutions. Let us first define the problem more formally. We start with a given set \( S \) of equations of the following form:

\[
S: \begin{align*}
R_1 &= E_1(R_1, \ldots, R_n) \\
R_2 &= E_2(R_1, \ldots, R_n) \\
& \cdots \\
R_n &= E_n(R_1, \ldots, R_n)
\end{align*}
\]

where each \( R_i \) is a distinct relational variable and each \( E_i \) is an expression of \( RA \), which involves some (not necessarily all) variables \( R_1, \ldots, R_n \); we call \( E_i \) the defining part of \( R_i \).

A substitution consists in the replacement of some variable \( R_i \) by its defining part \( E_i \) in the RHSs of all equations of \( S \). A resolution of \( R_j \) consists of a series of substitutions which generate an equation \( R_j = E' \) that does not contain any variable different from \( R_j \). Clearly, if \( R_j \) has a resolution, then its value can be computed either by evaluation of \( E' \) or by the application of the closure operator to \( E' \), yielding \( \rho(R_j) \).

After successful resolution, \( R_j \) can be marked as a known constant relation and we can eliminate the defining equation for \( R_j \) from the original system of equations \( S \). This step is called the elimination of \( R_j \) from \( S \). A set of equations is reducible by substitutions iff it can be transformed to the empty set by successive resolutions and eliminations. Note that it is useless to substitute a variable which occurs in its own defining part; we shall forbid such substitutions.

It is clear that the particular form of each expression \( E_i \) in \( S \) as well as the constant relations that appear in \( E_i \) are not relevant for determining whether \( S \) is reducible by substitutions; the only relevant information is the mutual interdependency of relational variables in \( S \). The information on variable interdependency is most appropriately represented in the dependency graph \( G(<N,E>) \) defined as follows:

\[
G: \begin{align*}
R_1 &= E_1(R_1, \ldots, R_n) \\
E(G) &= \{E_1, E_2, \ldots, E_n\} (E_i \text{ occurs in } E_i) \\
\text{Note that loops of the form } E_i E_j \text{ may occur in } E(G) \text{; in this case we call } R_i \text{ a looping node.}
\end{align*}
\]

It is easy to see that substituting a variable \( R_i \) in \( S \) exactly corresponds to dropping the node \( R_i \) from \( G \) and linking all the predecessors of \( R_i \) to all successors of \( R_i \) by new edges. By analogy, we call such a process the substitution of node \( R_i \) in \( G \). Since we have forbidden to substitute variables \( R_i \) that occur in their own defining part, we forbid to substitute looping nodes in a dependency graph.

\( L \), after a series of node substitutions, the node \( R_j \) has no outgoing edges to any nodes different from \( R_j \), we say that we have resolved \( R_j \). After a successful resolution of \( R_j \) we can eliminate node \( R_j \) and all its incident and outgoing edges from the original graph \( G \), yielding a new (and simpler) graph \( G' = G - C(R_j) \). If, by successive resolution and elimination of nodes, it is possible to get an empty graph, we say that \( G \) is reducible by substitutions.

**Lemma 3.1.** Any acyclic connection graph \( G \) is reducible by substitutions.

**Theorem 3.1.** A strongly connected graph \( G \) is reducible by substitution iff it contains a node \( K \) such that \( G - (K) \) is acyclic.

**Theorem 3.2.** A graph \( G \) is reducible by substitution iff its strongly connected components are each reducible by substitutions.

Proofs of Lemma 3.1 and of Theorems 3.1 and 3.2 are presented in [Camerini86]. Based on Theorems 3.1 and 3.2, we build algorithm REDUCE which determines whether a graph \( G \) is reducible by substitutions; in the positive case, the algorithm outputs a series of substitutions and resolutions (in correct order) to reduce \( G \).

The algorithm requires introducing a Reduction Graph \( \bar{G}(N,E) \), built from the dependency graph \( G \) by identifying all strongly connected components of \( G \). \( E(G) \) are the edges between connected components of \( G \), defined in the obvious way.

Clearly \( G \) is acyclic; we call bottom nodes of \( G \) all nodes which have no outgoing edges.

**ALGORITHM "REDUCE"
**

**INPUT:** Dependency Graph \( G \)

**OUTPUT:** Sequence of substitutions or resolutions

1. identify strong connection components \( C_1, \ldots, C_k \) of \( G \).

2. for each component \( C_i \) find a node \( K_i \) in \( C_i \) such that \( C_i - (K_i) \) is acyclic; if \( K_i \) cannot be found then stop with output "irreducible".

3. build the reduction graph \( \bar{G} \).

4. if \( \bar{G} \) is empty then stop.

5. for any bottom node \( C_i \) of \( \bar{G} \) do:
   for each node \( R_j \) in \( C_i - (K_i) \)
   output "substitute \( R_j \)";
   output "resolve \( K_i \)";
   apply REDUCE to \( C_i - (K_i) \);
   remove \( C_i \) from \( \bar{G} \).

6. goto 4.

It is easy to see that the REDUCE algorithm is polynomial in the size of its input (i.e., the dependency graph \( G \)); the identification of the strong connection components (step 1) as well as the acyclicity test (step 2) are well known polynomial problems; furthermore it is not hard to see that steps 4-6 are repeated at most \(||\bar{G}||\) times.

**Example.** Consider the following system \( S \) of equations in \( RA \):

\[
S: \begin{align*}
R_1 &= C_1 U R_1 (R_1 \bowtie R_6) U R_5 \\
R_2 &= R_5 U R_4 \\
R_3 &= T_{12} (R_4 \bowtie R_3) U R_2 \\
R_4 &= R_3 \bowtie R_3 \\
R_5 &= T_{13} (R_1 \bowtie R_6) U R_3 \\
R_6 &= R_1 \bowtie R_5 U C_5 \\
\end{align*}
\]

If, by successive resolution and elimination of nodes, it is possible to get an empty graph, we say that \( G \) is reducible by substitutions.

**Lemma 3.1.** Any acyclic connection graph \( G \) is reducible by substitutions.

**Theorem 3.1.** A strongly connected graph \( G \) is reducible by substitution iff it contains a node \( K \) such that \( G - (K) \) is acyclic.

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The algorithm requires introducing a Reduction Graph \( \bar{G}(N,E) \), built from the dependency graph \( G \) by identifying all strongly connected components of \( G \). \( E(G) \) are the edges between connected components of \( G \), defined in the obvious way.

Clearly \( G \) is acyclic; we call bottom nodes of \( G \) all nodes which have no outgoing edges.

**ALGORITHM "REDUCE"
**

**INPUT:** Dependency Graph \( G \)

**OUTPUT:** Sequence of substitutions or resolutions

1. identify strong connection components \( C_1, \ldots, C_k \) of \( G \).

2. for each component \( C_i \) find a node \( K_i \) in \( C_i \) such that \( C_i - (K_i) \) is acyclic; if \( K_i \) cannot be found then stop with output "irreducible".

3. build the reduction graph \( \bar{G} \).

4. if \( \bar{G} \) is empty then stop.

5. for any bottom node \( C_i \) of \( \bar{G} \) do:
   for each node \( R_j \) in \( C_i - (K_i) \)
   output "substitute \( R_j \)";
   output "resolve \( K_i \)";
   apply REDUCE to \( C_i - (K_i) \);
   remove \( C_i \) from \( \bar{G} \).

6. goto 4.

It is easy to see that the REDUCE algorithm is polynomial in the size of its input (i.e., the dependency graph \( G \)); the identification of the strong connection components (step 1) as well as the acyclicity test (step 2) are well known polynomial problems; furthermore it is not hard to see that steps 4-6 are repeated at most \(||\bar{G}||\) times.
Figure 1 shows the reduction graph for $S$.

Fig. 1. Dependency graph and reduction graph for a system of equations

The output produced by the REDUCE algorithm is as follows:

**SUBSTITUTE** $R_4$; **RESOLVE** $R_3$; **RESOLVE** $R_4$;
**SUBSTITUTE** $R_5$; **RESOLVE** $R_6$; **RESOLVE** $R_1$;
**RESOLVE** $R_6$; **RESOLVE** $R_5$; **RESOLVE** $R_2$.

The sequence of solutions is produced as follows:

- $R_3 = \left< R_3 \Rightarrow R_4 \Rightarrow R_5 \Rightarrow R_6 \Rightarrow R_1 \Rightarrow R_3 \right>$
- $R_4 = \left< R_4 \Rightarrow R_5 \Rightarrow R_6 \Rightarrow R_1 \Rightarrow R_3 \right>$
- $R_5 = \left< R_5 \Rightarrow R_6 \Rightarrow R_1 \Rightarrow R_3 \right>$
- $R_6 = \left< R_6 \Rightarrow R_1 \Rightarrow R_3 \right>$

3.5. Resolution of mutually dependent equations

Some systems of equations cannot be reduced by substitutions; in such cases we have to use other solution methods. There are two approaches for solving nonreducible systems. In the first approach, the single relation variables $R_i$ are combined to one super-variable $R$, for instance by use of cartesian products. Then the original system of equations can be rewritten as $R = E[R]$ and solved by applying the closure operator. Two different variants of this method are described in [Chandra82] and [Ceri86].

The second approach computes the solution directly from the original system of equations by initially setting $R_i = \phi$ for $i = 1 \ldots n$ and then successively computing $R_i = E[R_1 \ldots R_n]$, until the value of each $R_i$ remains unchanged.

**ALGORITHM A**

```plaintext
FOR i := 1 TO n DO Ri := \phi;
REPEAT
  cond := true;
  FOR i := 1 TO n DO Si := Ri;
  FOR i := 1 TO n DO BEGIN
    Ri := Ei[S1, \ldots, Sn];
    IF Ri \neq S1 THEN cond := false;
  END;
UNTIL cond;
FOR i := 1 TO n DO OUTPUT(Ri).
```

We can optimize Algorithm A by using at each step the recently computed values for $R_1, \ldots, R_i$ for the computation of the new value of $R_{i+1}$, instead of using the old values, i.e., $S_1, \ldots, S_i$:

**ALGORITHM B**

```plaintext
FOR i := 1 TO n DO Ri := \phi;
REPEAT
  cond := true;
  FOR i := 1 TO n DO BEGIN
    S := Ri;
    R1 := E1[R1, \ldots, Rn];
    IF Ri \neq S THEN cond := false;
  END;
UNTIL cond;
FOR i := 1 TO n DO OUTPUT(Ri).
```

Algorithms A and B have well-known correspondents in the field of numerical analysis: Algorithm A corresponds to the Jacobi algorithm for the iterative solution of systems of equations, while algorithm B corresponds to the Gauss-Seidel algorithm.

3.6. Conclusion of Section 3

In this section we have given rules for transforming FFLPs into systems of equations in RA+ and we have shown how these systems can be solved by use of the closure operator. By this process we have defined both the fixpoint semantics and the operational semantics for FFLPs and queries operating on an extensional database EDB.

Though outside the scope of this paper, it can be seen that our semantics correspond exactly to the semantics for logic programs defined by Van Emden and Kovalski [VanEmden76]; due to the limitedness of ranges for individual variables and to the absence of function symbols, the Herbrand universe of any FLP is finite.

4. ALGEBRAIC APPROACH TO THE OPTIMIZATION OF LOGIC PROGRAMS

We turn now to the optimization of expressions in ERA+. Execution strategies presented in the previous section suffer from two major disadvantages:

a. The algorithm which computes the closure operation is not too efficient.
b. In the computation, $O(E_1 E_2)$, conditions of the logical query are not used, because the selection condition is not pushed inside $0 E_1 E_2$.

Proofs of theorems in this section are quite simple and can be found in [Ceri86].

4.1. Efficient computation of the closure operator

Algorithm $A_0$ presented in section 2.2 is not very efficient because several partial unions are needed and because $E(S)$ has to be evaluated several times for the same tuples; consider now the more efficient program $A_1$:

**$A_1(E)(X)$**
Theorem 4.1. If $E(X)$ is an expression in $RA^+$, then $A_0(E(X))$ and $A_1(E(X))$ are equivalent programs.

Note that Theorem 4.1 does not hold if $E(X)$ is not in $RA^+$; for instance, if $E(X)=K-X$, $K \neq \emptyset$, then $A_1$ never halts while $A_0$ terminates returning $K$ as result.

Program $A_1$ eliminates the first source of inefficiency, i.e. the computation of unions within the program $A_0$, but it does not eliminate the second source, i.e. the computation of $E(S)$ several times for the same tuple. We now develop two algorithms which do not have this burdensome property.

Definition 4.1. An expression $E(X)$ over $RA^+$ is linear if it holds:

$$\forall X, Y \in \mathcal{R}_g(E) : E(X \cup Y) = E(X) \cup E(Y)$$

We can now present algorithm $A_2$:

$$A_2(E(X)) :$$

```plaintext
S <- \emptyset
REPEAT
Y <- S
S <- E(S)
UNTIL Y = S
RETURN S
```

Theorem 4.2. If $E(X)$ is linear, then $A_2(E(X))$ is equivalent to $A_0(E(X))$.

The advantage of algorithm $A_2$ compared to algorithm $A_1$ is that the size of $D_i$ (the "difference term" produced at each iteration) is smaller with respect to the size of $S_i$ (the "accumulation" term).

Theorem 4.3. Each expression $E(X)$ can be rewritten in a canonical form:

$$U_{1=1} \prod_{c_i}^n D_i = \bigcup_{c_i}^n (X^{Y_i} \times C_i) \cup C_0$$

where $X^{Y_i}$ represents the cartesian product of $y_i$ terms and $C_i$ are constant (database) relations.

Let $\text{can}(E(X))$ denote the transformation in $RA^+$ from an expression $E(X)$ to its canonical form. Let $\text{degree}(E(X))$ denote the maximal $i$ in $\text{can}(E(X))$.

Algorithm $A_2$ can be applied.

Let $E'(X,Y)$ be the expression obtained from $E(X)$ by replacing each cartesian product $X \times X$ with $X \times Y$. Then, we can build algorithm $A_3$.

$$A_3(E(X)) :$$

```plaintext
S <- \emptyset, D <- \emptyset
REPEAT
T1 <- E'(S,D)
S <- T1
D <- (T1 \cup T2 \cup T3) - S
S <- S \cup D
UNTIL D = \emptyset
RETURN S
```

Theorem 4.4. If the degree of $E(X)$ is 2, then algorithm $A_3$ is equivalent to $A_1$.

The advantage of algorithm $A_2$ compared to algorithm $A_1$ is that we never compute the term $E'(S,S)$, which might be large. We further notice that the program can be simplified by omitting the evaluation of $T_2$ if the expression $E'(X,Y)$ is commutative. Algorithm $A_3$ can be generalized for a generic $i$.

Example: Nonlinear ancestor program

$$\text{FFLP:} \text{anc}(x,y) :- \text{par}(x,y).$$
$$\text{anc}(x,y) :- \text{anc}(x,z), \text{anc}(z,y).$$

\text{DISEQUATIONS:} \text{PAR} \subseteq \text{ANC}$$

$$\Pi \subseteq \Pi \quad (\text{ANC} \bowtie \subseteq \text{ANC}) \subseteq \text{ANC}$$

\text{EQUATION:} \text{ANC} = \text{PAR} \cup (\text{ANC} \bowtie \subseteq \text{ANC})$$

\text{SOLUTION:} \text{sol} \subseteq \text{ANC} = (\Pi \subseteq \Pi \bowtie \subseteq \text{ANC}) \cup (\text{ANC} \bowtie \subseteq \text{ANC})$$

The degree of $E(\text{ANC})$ is 2. Assuming acyclicity of the \text{PAR relation}, algorithm $A_2$ produces at each iteration $i$ the pair of ancestors corresponding to the $2i$-th and $2i+1$-th generations; term $T_2$ should not be evaluated, as the expression is clearly commutative. Acyclicity of the \text{PAR relation is not required by algorithm $A_3$.}

4.2. Pushing selection conditions into linear expressions

Aho and Ullman [Aho79] indicate a method for optimizing expressions of the type: $\Phi \bowtie E(X)$. We briefly outline their method by one example. The linear expression for the set of all ancestors of an individual "$a$" is:

$$G_1 \bowtie \text{ANC} = \text{ANC} = \emptyset$$

It holds:

$$G_1 \bowtie \text{ANC} = \text{ANC} = \emptyset$$

By applying associativity and distributivity:

$$G_1 \bowtie \text{ANC} = \emptyset$$

By introducing the variable $Y$ for $G_1 \bowtie \text{ANC}$ we get:

$$\text{ANC} = \emptyset$$

This formula can now be evaluated using algorithm $A_3$. Unfortunately, this method applies only when the selection can be pushed directly to the variable $X$ in $E(X)$. In the rest of this section we show, on the ground of examples, how other optimizations are possible. Consider the terms:

$$\text{(4.1) } D_0 = E(\Phi)$$
$$\text{(4.2) } D_{n+1} = E(D_n) - E(\Phi)$$

Considering algorithm $A_2$, it is easy to verify that, for linear expressions $E(X)$, the following equation holds:

$$\Phi = \bigcup_{i=0}^{n} D_i$$

This formulation for $E(X)$ is attractive because we can compute terms so that $\Phi$ is pushed into each $D_i$ and never compute the full terms $D_i$.

$$\text{(4.2) } G_1 \bowtie E(X) = \bigcup_{i=0}^{n} G_1 \bowtie D_i$$
Starting from this general form, we can see how algebraic manipulations produce the same effect as techniques such as the "magic set" and "magic counting". We use a well-known example: the search of same generation cousins. Let us produce the transformation from FFLP to ERA+ for this example:

FFLP: \( \text{sg}(X,Y) \).

\[ \text{sg}(X,Y) = \text{par}(X,X_1), \text{sg}(X_1,Y_1), \text{par}(Y,Y_1). \]

\begin{align*}
\text{UISEQUALATIONS: } & \text{EQ } \land \text{SG} \\
& \left( \text{par} \land \text{sg} \right) \land \left( \text{par} \land \text{sg} \right) \\
\text{EQUATION: } & \text{SG} = \text{EQ UT} \\
& \text{par} \land \text{sg} \\
\text{SOLUTION: } & \text{sol(SG) = 0} \\
& \left( \text{par} \land \text{sg} \right) \land \left( \text{par} \land \text{sg} \right) \\
\end{align*}

Notice that the expression computing SG is linear.

We introduce the "composition" operation, as in [Aho 79], to denote the following expression (where \( R \) and \( S \) are binary relations):

\[ R \circ S = \{ (r, s) : \exists t \in S \text{ such that } (r, t) \in R \text{ and } (t, s) \in S \}. \]

\[ \text{R o S = T} \]

Then, denoting as \( RAP \) the relation obtained by exchanging the order of attributes in \( PAR = T \), we have:

\[ \text{sol(SG) = 0} \]

\[ \text{EQ UT} \]

Note that the composition operation is associative, hence:

\[ (X \circ Y) \circ Z = X \circ (Y \circ Z) = X \circ Y \circ Z \]

Furthermore, we indicate as \( X_i \) a chain of \( i-1 \) applications of the composition to a binary relation \( X \):

\[ X_i = X_1 \circ X_2 \circ \ldots \circ X_i \]

Terms \( Di \) defined by the system (4.1) are:

\[ D_i = \text{EQ} \]

\[ D_{i+1} = \text{EQ} \]

Let us compare the terms \( D_i \) and \( D_{i+1} \): we denote as \( R_i \) the largest common subexpression(s) of \( D_i, D_{i+1} \) which includes selection condition(s); in this case, \( R_i = \text{EQ} \).

It is possible to pre-determine a subset of the relation \( PAR \) which contains all relevant tuples for the computation of \( SG \); this is done by evaluating the "magic" set \( M \) of all elements that can appear in the second column of terms \( R_i \); this is the set of all ancestors of "a".

\[ \text{PAR}_M = \phi \]

\[ i = 0 \]

while \( R_i \) is not empty do

\[ \text{PAR}_M = \text{PAR}_M \cup R_i \]

\[ i = i + 1 \]

end

\[ M = \text{PAR}_M \]

Consider now the semi-join reduction of \( PAR \):

\[ \text{PAR'} = \text{PAR} \circ \text{PAR} \]

It is easy to see that only tuples of this relation give a contribution to terms \( D_i \); we can then write:

\[ D_i = \text{SG} \circ \text{PAR'} \circ \text{PAR} \]

By using equation (4.2), we deduce that:

\[ G_{1-i-a} \circ \text{SG} \circ \text{PAR'} = \text{PAR} \circ \text{PAR} \circ \text{PAR} \]

We can then apply algorithm \( A_2 \) to solve this simplified problem. Note that \( M \) is itself obtained as the application of the closure to a simple expression, as follows:

\[ M = \text{PAR} \]

Assuming acyclic data, we can easily show the algebraic equivalent of the "magic counting" method. Let us first simplify each term \( D_i \) by eliminating the \( EQ \) relation and propagating equality conditions:

\[ D_i = \text{PAR'} \circ \text{RAP} \]

With cyclic data, there cannot be replicated tuples in the union of terms \( R_i \); hence, there is an \( i \) such that \( R_i = \phi \) for \( i > 1 \). But if \( R_i = \phi \), then \( D_i = \phi \). Hence we can evaluate \( SG \) using the LHS of equation (4.2) by the following algorithm:

\[ \text{SG} = \phi \]

\[ R = \text{PAR} \]

while \( R \neq \phi \) do

\[ \text{begin} \]

\[ R = R \circ \text{PAR} \]

\[ \text{SG} = \text{SG} \cup (R \circ \text{PAR}) \]

\[ i = i + 1 \]

end

Algebraic transformations of this section are easily generalized for a query with two bindings:

\[ \text{-sg(a,b)} \]

corresponding to the expression:

\[ \text{SG} = \text{PAR} \]

We omit details of derivations, and show the final results:

1) Using magic sets:

\[ \text{PAR} \]

2) Using magic counting, we produce the program:
Using magic counting, we produce the program:

\[
\begin{align*}
\text{answer}=\text{no} \\
R_1 &= \emptyset \text{ PAR} \\
R_2 &= \emptyset \text{ PAR} \\
\text{if } R_1 \circ R_2 \neq \emptyset \text{ then answer }=\text{yes} \\
\text{while } ((R_1 \neq \emptyset) \text{ and } (R_2 \neq \emptyset) \text{ and } (\text{answer }= \text{no})) \text{ do} \\
& \quad \begin{align*}
R_1 &= R_1 \cup \text{ PAR} \\
R_2 &= \text{ PAR} \cup R_2 \\
\text{if } R_1 \circ R_2 \neq \emptyset \text{ then answer }=\text{yes} \\
\end{align*} \\
\text{end} \\
\text{output answer}
\end{align*}
\]

Notice that with this program we compute iteratively two terms, each obtained from one binding condition. The computation is halted as soon as either of the two terms is empty or \( D_i = R_1 \circ R_2 \) produces one tuple.

5. CONCLUSIONS

This paper has presented a systematic approach to the algebraic treatment of logic queries; we have shown a syntax directed translation from FFLP to algebraic equations and then shown how equations or systems of equations can be solved and how individual equations can be optimized.

Several problems considered in this paper need further improvements:

a. The proposed solution method for systems of equations could be improved by propagating bindings from one equation to another.

b. Efficient algorithms presented for expressions of degree 1 and 2 can be generalized to expressions of any degree.

c. Further investigation is needed to fully understand how the algebraic approach compares with the "magic set" and "magic counting" methods.

d. Another noticeable direction of research has as goal the treatment of Horn clauses including function symbols. The necessary counterpart on the database side is the extension of the relational model and languages to model complex objects (e.g., non-INF relations).

Acknowledgement

This research was supported by the Eerist project n.432 Meteor (An integrated formal approach to industrial software development). We like to thank Stefano Crespi-Reghizzi, who has stimulated our work and provided useful comments on a first draft of the paper, Letizia Tanca, who has suggested us an improvement for the solution of systems of equations, and Paolo Camerini, who has helped proving the theorems in Section 3.4.

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