

DATA CONSTRUCTORS: ON THE INTEGRATION  
OF RULES AND RELATIONS

Matthias Jarke

Graduate School of Business Administration  
New York University

Volker Linnemann and Joachim W. Schmidt

Fachbereich Informatik, Johann Wolfgang Goethe-Universität  
Dantester. 9, 6000 Frankfurt 1, West Germany

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Center for Research on Information Systems  
Computer Applications and Information Systems Area  
Graduate School of Business Administration  
New York University

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## **Data Constructors: On the Integration of Rules and Relations**

### **Abstract**

Although the goals and means of rule-based and data-based systems are too different to be fully integrated at the present time, it seems appropriate to investigate a closer integration of language constructs and a better cooperation of execution models for both kinds of approaches.

In this paper, we propose a new language construct called **constructor** that - when applied to a base relation - causes relation membership to become true for all tuples constructable through the predicates provided by the constructor definition. The approach is shown to provide expressive power at least equivalent to PROLOG's declarative semantics while blending well both with a strongly typed modular programming language and with a relational calculus query formalism. A three-step compilation, optimization, and evaluation methodology for expressions with constructed relations is described that integrates constructors with the surrounding database programming environment. In particular, many recursive queries can be evaluated more efficiently within the set-construction framework of database systems than with proof-oriented methods typical for a rule-based approach.

### **1 Introduction**

Combining the semantic capabilities of rule-based knowledge representation and reasoning systems with the efficiency-oriented mechanisms for query result construction and transaction processing in large shared DBMS has been the focus of much recent research [Kers 84]. Apart from the possibility of defining a completely new architecture for "knowledge based systems", the solutions proposed so far can be interpreted as extreme points in a continuum of coupling strategies. Researchers either propose to replace one system completely by the other one [ScWa 84] - the end points of the spectrum - or to couple current expert systems languages (most notably, PROLOG [JaVa 84], [Zani 84]) with existing DBMS interfaces - the cutting point defined by history.

With a number of researchers [Smith 84], [Ullm 84] we believe that a coupling strategy is preferable to fully integrated solutions. Because of the different stress on representation vs. efficiency between KR and DB research [MyBr 85], little is gained (and unnecessary complexity is incurred) by putting all capabilities into one system. In contrast, a coupling architecture allows each subsystem to evolve independently and to offload the reconciliation task to separate coupling tools [Jark 84].

Granted that coupling is necessary, the question remains what the capabilities of each of the partners should be. While in the short run there is a clear economic incentive to leave existing systems as they are [JaVa 84], nothing says that the optimal division of labour between the deductive capabilities of rule-based systems and the selective power of data-based systems will remain where it has been historically - at the point of 'relational completeness' as defined by [Codd 72].

The question of exactly what capabilities should be added to the DBMS is open. [Ullm 84] proposes an architecture using 'capture rules' that define useful extended DBMS capabilities. The present paper follows a similar approach but investigates the problem from the viewpoint of integrated database programming languages rather than from a PROLOG perspective. Database languages can in principle handle, e.g., recursive queries using programming language constructs, such as functions and iteration. However, current query optimization strategies do not take advantage of the relationships among the corresponding sequence of queries. Efficiency is the responsibility of the programmer.

Rather than enhancing a query optimizer directly to handle multiple related queries, this paper studies special-purpose language constructs that capture higher-level data definition and operation and are easily recognizable by a compiler. To provide the necessary framework, section 2 reviews the database programming language DBPL which integrates relational data structures and transactions with the programming language MODULA-2 [Wirth 83].

The main focus of the paper is the detailed analysis of a DBPL extension called **constructor** which has evolved from the **selector** concept introduced in [MaReSc 84]. While selectors allow the definition of selected subrelation variables, constructors expand existing relations. Constructors implement recursion using an equational fixed point semantics. We introduce constructors in sections 2 and 3 using an example with a CAD flavour. We show how to integrate the tuple relational calculus concepts of negation and universal quantification into this framework. Moreover, we demonstrate that our proposal provides expressive power at least equivalent to PROLOG's clause-order independent declarative semantics while being closer to the spirit of typed, procedural database programming languages, such as PASCAL/R [Schm 77] or ADAPLEX [Smith 81].

The database programming language environment also inspires particular implementation and optimization strategies since it is frequently used for implementing higher-level database interfaces. In section 4, we interpret constructed relations as an extension to range-nested expressions as introduced in [JaKo 83], and outline a three-level compilation and optimization framework.

## 2 Types, Relations, and Predicates

The impact of logic on computing - from early data processing in the fifties to modern computer science - can hardly be overestimated.

In the field of programming, logic marks the step from machine-oriented coding to algorithmic programming. High level languages provide conditional statements and boolean expressions, use propositions for data type definition, and depend crucially on predicates for the specification of language semantics and for reasoning about programs [Gries 81], [Hehn 84].

In the area of data modelling, the degree to which predicates are utilized allows a distinction between early reference-oriented data models and those that capture more of the relationships defined by the application semantics.

### 2.1 Data Types and Predicates

If "a type is a precise characterization of structural and behavioural properties which a collection of entities (actual or potential) all share ..." [Deut 81], the formalism by which those properties can be characterized decides upon the power of a type calculus.

Currently prevalent programming languages only allow type definitions based on restricted propositional logic. Take, as an example, the following PASCAL-like subtype definition:

```
partidtype IS RANGE 1..100,
```

which is equivalent to the domain predicate ( $1 \leq p$  AND  $p \leq 100$ ) and defines the domain set

```
partidtype ( EACH p IN integer: 1 <= p AND p <= 100 ) .
```

The expressiveness of the type calculus in high level languages corresponds closely with that of the expression and statement part of these languages. As a consequence, any action to be taken to assure type properties can be expressed directly in the language. A type checker can produce run time code in the source language to assure, for example, type correctness of an integer expression,  $ix$ , which is to be assigned to a variable,  $p$ , of partidtype:

```
IF (1<=ix) AND (ix<=100)
THEN p:=ix
ELSE <exception> .
```

Programmers reduce the possibility of run-time exceptions by acquiring sufficient information on rhs-expressions through inductive reasoning about assignment chains and subtype definitions (and so do clever compilers).

Approaches to programming that are more concerned about correctness allow for the definition of additional program properties by so-called annotations. ADA annotations, for example, can be specified in the meta language ANNA [KrBr 84], and ADA programs can be proven formally correct w.r.t. their specification. The meta language ANNA allows full first order assertions, while the object language ADA is restricted to propositional logic. An ADA subtype definition, for example, primetype, can be fully specified by the following ANNA annotation [KrBr 84]:

```
primetype IS integer || WHERE p IN primetype ==>
ALL n IN integer
((1<n AND n<p) ==> p MOD n # 0),
```

defining the domain set

```
primetype ( EACH p IN integer: ALL n IN integer
            ((1<n AND n<p) ==> p MOD n # 0) ) .
```

## 2.2 Predicates in Database Languages

Database models, as, for example, the relational model are very concerned about data integrity; therefore they go beyond programming languages in the sense that they provide the expressiveness of first order logic directly through relational languages.

On the expression level, the request for "relational completeness" of query languages is essentially met by allowing full first order predicates,  $p(r, \dots)$ , as selection predicates in relational expressions:

```
reltype (EACH r IN rel: p(r, \dots)).
```

On the type or schema level, the role of predicates can be exemplified best by comparing a PASCAL-like set type definition

```
settype = SET OF elementtype.
```

with a relation type definition.

The legal values of a relation are also sets of elements; they have to meet, however, the additional constraint that some attribute (or a collection of attributes) serves as a key, i.e., has a unique value amongst all the elements of a relation:

```
reltype = SET OF elementtype ||
          WHERE rel IN reltype ==>
          ALL r1,r2 IN rel ( r1.key=r2.key ==> r1=r2 ) .
```

The key constraint is essential to relational data modelling since only unique keys can serve as element identifiers as required, for example, for the construction of higher relationships between elements. Therefore, relational languages directly support the above class of annotated set type definitions by a data structure **relation** that allows for type definitions equivalent to the previous one:

```
reltype = RELATION key OF elementtype.
```

For each assignment of a relational expression,  $rex$ , to a variable,  $rel$ , of  $reltype$ , the relational type checker has to perform a test equivalent to

```
IF ALL x1,x2 IN rex ( x1.key=x2.key ==> x1=x2 )
THEN rel:=rex
ELSE <exception> .
```

## 2.3 Predicative Support for Relations: Selectors and Constructors

The key constraint is, of course, not the only condition one would like to have maintained automatically on a database. Take, for an example, some objects related by the fact that one object is in front of another.

```
TYPE objecttype = ... (* full object description,
                       e.g. by object record *) ... ;
parttype = ... (* representative object description,
                e.g. by object key *) ... ;
```

```

objectrel = RELATION part OF objecttype;
infrontrel = RELATION ... OF
              RECORD
              front,back: parttype
            END;
VAR Objects: objectrel;
    Infront: infrontrel .

```

Since the attributes, front and back, of the Infront relation are supposed to relate objects, they have to refer to elements in the Objects relation. The corresponding referential integrity constraint can be expressed by annotating the type of the Infront relation:

```

VAR Infront: infrontrel || WHERE r IN Infront ==>
    SOME r1,r2 IN Objects
    (r.front=r1.part AND
     r.back=r2.part).

```

In a relational language such a constraint can be enforced by a conditional which controls assignment to the Infront relation:

```

IF ALL x IN rex (SOME r1,r2 IN Objects
    (x.front=r1.part AND x.back=r2.part) )
THEN Infront:=rex
ELSE <exception> .

```

Expecting frequent use of relations in such "conditional patterns", the database programming language DBPL [ScMa 83], [MaReSc 84]] provides an abstraction mechanism for such patterns through the notion of a **selector**. Referential integrity on relations of type infrontrel, for example, can be maintained by

```

SELECTOR refint FOR Rel: infrontrel(): ...;
BEGIN EACH r IN Rel: SOME r1,r2 IN Objects
    (r.front=r1.part AND r.back=r2.part)
END refint .

```

An assignment to a selected relation variable, for example,  
 Infront[refint] := rex,

is defined to be equivalent to the above conditional assignment to the full relation variable Infront.

In summary, selectors "factor out" conditions on relations, represent them uniformly, and make them available to all database system components that have to reason about programs and data (such as query optimizer, concurrency manager, and integrity subsystem). The selector concept is illustrated in Fig. 1.

While selectors provide support when data elements are to be **excluded** from a relation there is also a need for supporting the contrary - when additional derived data objects are to be **included** into a relation.

For an example, a relation, Ahead\_2, can be defined that relates - based on the data in relation Infront - two objects if and only if they are separated by at most two steps.

Starting with type

```

TYPE aheadrel = RELATION ... OF RECORD
    head,tail: parttype
  END.

```

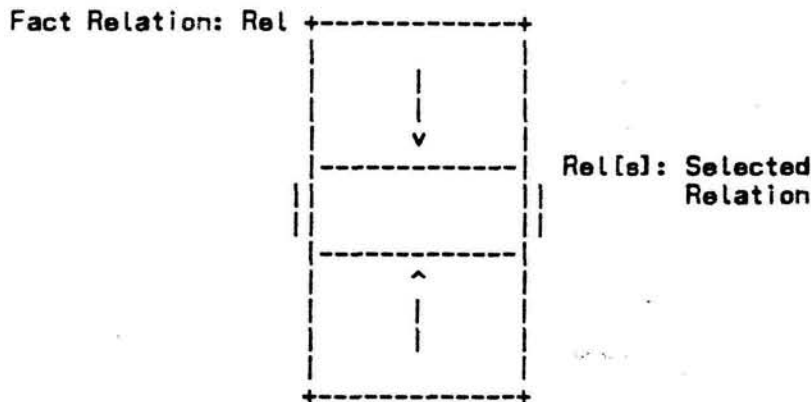


Figure 1: Selectors and Relations

an annotated definition of relation Ahead\_2 would read as follows:

```
VAR Ahead_2: aheadrel || WHERE (r IN Infront ==> r IN Ahead_2)
AND (f,b IN Infront ==>
(f.back=b.front ==>
<f.front,b.back> IN Ahead_2)) .
```

In a relational language the value of such a relation, Ahead\_2, can be denoted by a query expression in terms of predicates and the Infront relation:

```
aheadrel { EACH r IN Infront: TRUE,
<f.front,b.back> OF EACH f,b IN Infront:
f.back=b.front }.
```

Expecting frequent use of relations in such "expressional patterns" this paper proposes an abstraction mechanism for such patterns based on the notion of a **constructor**.

As an example, the Ahead\_2-relationship based on relations of type infrontrel can be constructed by

```
CONSTRUCTOR ahead_2 FOR Rel:infrontrel (): aheadrel;
BEGIN EACH r IN Rel: TRUE,
<f.front,b.back> OF
EACH f,b IN Rel: f.back=b.front
END ahead_2.
```

The value of a constructed variable, for example,

```
Infront {ahead_2}
```

is defined to be equal to the above relational expression of type aheadrel.

In the same sense as selectors isolate the constraints imposed on selected relations, constructors factor out the rules that define the elements in constructed relations.

In the subsequent section, the basic issues of constructor semantics are discussed with emphasis on recursive constructor definition and constructor convergence, and constructors are compared with other approaches to rule and fact management.

### 3 Relation Constructors

Many database rules follow a similar pattern. Certain base facts are stored for which it is known that a certain rule holds. Other facts for which the rule also holds are not stored explicitly but can be derived by a possibly recursive (deduction) rule. The deduction rule may depend on the existence of other facts (parameters), which are, however, not necessarily part of the result. Constructors allow the definition of such deduction rules in DBPL. The idea is illustrated in Fig. 2.

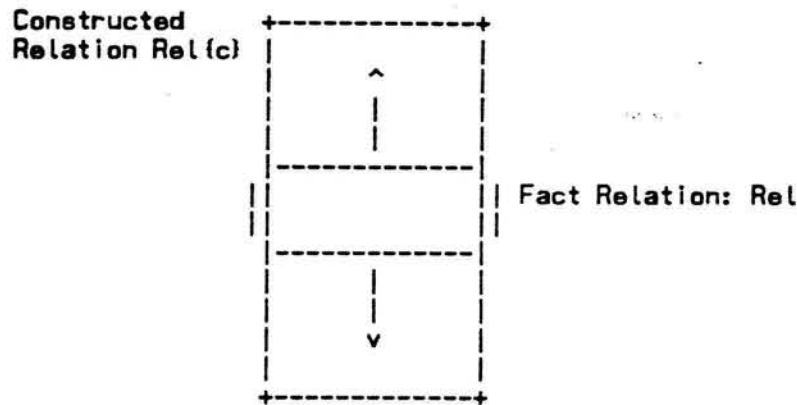


Figure 2: Constructor and Relations

As a simple example, the transitive closure of a relation contains the relation itself and other tuples derived from it.

In this section, we discuss the notion of a constructor in more detail. We first provide some examples based on the relations introduced in section 2, and then define the semantics of recursive constructors formally.

#### 3.1 Recursive Constructors

The above simple constructor, ahead\_2, representing all object pairs separated by at most two steps, can be generalized to a sequence of constructors, ahead\_n, representing all pairs of objects separated by at most n steps:

```

CONSTRUCTOR ahead_n FOR Rel:infrontrel(): aheadrel;
BEGIN  EACH r IN Rel: TRUE,
      <f.front,b.tail> OF EACH f IN Rel,
      EACH b IN Rel (ahead_n-1) :
      (f.back=b.head)
END ahead_n .

```

For the definition of a constructor, ahead, representing all object pairs separated by an arbitrary number of steps, we utilize **simple recursion**:

```

CONSTRUCTOR ahead FOR Rel:infrontrel(): aheadrel;
BEGIN  EACH r IN Rel: TRUE,
      <f.front,b.tail> OF EACH f IN Rel,
      EACH b IN Rel (ahead):
      (f.back=b.head)
END ahead .

```



Intuitively, the value of a constructed relation

```
Infront {ahead}
```

can be seen as the limit of the sequence of constructor applications:

```
Infront {ahead} =  $\lim_{n \rightarrow \infty}$  Infront {ahead_n}
```

The details of constructor semantics are given in section 3.2.

Because, in our example, the sequence is monotonic, the limit exists and can be implemented by a finite loop using a relation variable, Ahead:

```
Ahead := ();
REPEAT
  Oldahead := Ahead;
  Ahead := (EACH r IN Infront:TRUE,
            <f.front,b.tail> OF EACH f IN Infront,
            EACH b IN Ahead:
            f.back=b.head )
UNTIL Ahead = Oldahead .
```

According to [AhUl 79], we compute the least fixed point of a relational expression.

To show how to combine selectors and constructors, we define

```
SELECTOR hidden_by (Obj: parttype) FOR Rel: infrontrel();
BEGIN
  EACH r IN Rel: r.front = Obj
END hidden_by .
```

Then the expression

```
Infront [hidden_by ("table")] {ahead}
```

returns all objects behind the table.

To give an example of **mutual recursion**, we introduce a second dimension to our object relationships. In addition to one object being in front of the other, objects may be related by the fact that one is on top of another. The new facts can be represented by the following relation:

```
TYPE ontoprel = RELATION ... OF
  RECORD
    top,base: parttype
  END;
VAR Ontop: ontoprel .
```

We say that object A is ahead of object B not only if it is (maybe indirectly) in front of object B but also if A is above an object that is (maybe indirectly) in front of B. For example, we would say that a vase is ahead of a chair if the vase is on top of a table which is in front of the chair. In order to reflect this extended relationship, we define a new data type

```
TYPE aboverel = RELATION ... OF
  RECORD
    high,low: parttype
  END
```

and introduce a second constructor:

```

CONSTRUCTOR above FOR Rel: ontoprel(Infront:infrontrel):
                                above:rel;
BEGIN  EACH r IN Rel: TRUE,
      <r.top,ab.low> OF
        EACH r IN Rel,
          EACH ab IN Rel {above(Infront)} :
            r.base = ab.high,
      <r.top,ah.tail> OF
        EACH r IN Rel,
          EACH ah IN Infront {ahead(Rel)} :
            r.base = ah.head
END above.

```

The constructor ahead is redefined to be mutually recursive, too:

```

CONSTRUCTOR ahead FOR Rel: infrontrel (Ontop:ontoprel):
                                ahead:rel;
BEGIN  EACH r IN Rel: TRUE,
      <r.front,ah.tail> OF
        EACH r IN Rel,
          EACH ah IN Rel {ahead(Ontop)} :
            r.back = ah.head,
      <r.front,ab.low> OF
        EACH r IN Rel,
          EACH ab IN Ontop {above(Rel)} :
            r.back = ab.high
END ahead .

```

Through these definitions we can combine both relations, Infront and Ontop, and both constructors, ahead and above, by, for example,

```

Infront {ahead(Ontop)}      and
Ontop {above(Infront)} .

```

The values of these mutually recursive constructed relations are defined by the limits of mutually defined sequences; again, the details are given in section 3.2.

Because the sequences are monotonic, the limits exists and can be implemented by the following loop, using auxiliary variables, Ahead and Above, for the values of the constructed relations:

```

Ahead:= {};  Above:= {};
REPEAT
  Oldahead:=Ahead;  Oldabove:=Above;
  Ahead:=ahead_fct(Oldahead,Oldabove);
  Above:=above_fct(Oldahead,Oldabove);
UNTIL Ahead=Oldahead AND Above=Oldabove.

```

ahead\_fct and above\_fct are relation-valued functions based on the definition of the constructors, ahead and above.

In most applications it is obvious to which selectors a constructor is to be applied (for example, Infront) and which relations are to serve as arguments (for example, Ontop). In a few cases, however, this choice may be difficult, and the programmer may prefer to start with an empty relation (for example, if the constructor is based on a join of several base relations rather than growing out of a single one).

### 3.2 Constructor Semantics

In general, a database program may contain a large number,  $m$ , of mutually dependent constructors:

```
CONSTRUCTOR  $c_1$  FOR Rel1: reltype1 (...): resulttype1;  
BEGIN  
   $f_1$  (... , applyc1,1, ..., applyc1,n1)  
END  $c_1$ ;  
...  
CONSTRUCTOR  $c_m$  FOR Relm: reltypem (...): resulttypem;  
BEGIN  
   $f_m$  (... , applycm,1, ..., applycm,nm)  
END  $c_m$  .
```

where each  $\text{apply}_{c_i,j}$  is a (possibly recursive) constructor application of the form  $\text{Rel } \{c(\dots)\}$ .  $\text{Rel}$  is a relation name known in the context of  $f_i$ , and  $c$  is one of our  $c_i$ .  $\{f_i\}$  is a relational calculus expression. To simplify indexing, we rename our constructor applications  $\text{apply}_{c_i,j}$  to  $\text{apply}_1, \dots, \text{apply}_l$ ,  $l = n_1 + \dots + n_m$ .

The semantics of a constructor application

$$\text{apply}_0 = \text{Actrel } \{ c_i (\dots) \},$$

on an actual relation  $\text{Actrel}$ , is defined as follows:

We construct  $l+1$  functions

$$g_0(\text{apply}_0, \text{apply}_1, \dots, \text{apply}_l)$$
$$\dots$$
$$g_l(\text{apply}_0, \text{apply}_1, \dots, \text{apply}_l);$$

function  $g_j$  is constructed by taking the function  $f_p$  which corresponds to the constructor in the application  $\text{apply}_j$ , and replacing all formal parameters by their actual values.

We define

$$\text{apply}_{i,0} = \{ \} \quad (i=0,1,\dots,l)$$
$$\text{apply}_{i,k+1} = g_i(\text{apply}_{0,k}, \dots, \text{apply}_{l,k})$$

and compute the limits:

$$\text{apply}_i = \lim_{k \rightarrow \infty} \text{apply}_{i,k}.$$

The value of constructor application  $\text{Actrel } \{ c_i (\dots) \}$  is given by  $\text{apply}_0$ .

Of course, this definition makes sense only if the limit of the above sequences exists. If the functions  $f_i$  are monotonic, we have  $\text{apply}_{i,0} \subset \text{apply}_{i,1}$ , and therefore, by induction,  $\text{apply}_{i,k} \subset \text{apply}_{i,k+1}$ . Because all relations are based on finite domains, there must be a step  $j$  such that  $\text{apply}_{i,j} = \text{apply}_{i,j+1}$ . Therefore, if the  $f_i$  are monotonic, the limits exist and are reached after finitely many steps. It can be shown [ChHa 82] that the functions  $f_i$  are monotonic if their predicates are free of negation and universal quantifiers.

Note that, according to [Tars 55], we compute the least fixed point of the system of equations

$$\text{apply}_0 = g_0(\text{apply}_0, \dots, \text{apply}_l)$$
$$\dots$$
$$\text{apply}_l = g_l(\text{apply}_0, \dots, \text{apply}_l).$$

A program for computing the limits can be written in the same way as for our examples in 3.1

### 3.3 Negation and Universal Quantification

Database languages like DBPL and Pascal/R [Schm 77] allow universal quantification of element variables as well as negation of relational predicates. However, constructors containing negation and universal quantification may be meaningless because the limit of the fixed point computation may not exist, as, for example, in

```
CONSTRUCTOR nonsense FOR Rel: anytype (): anyothertype;
BEGIN EACH r IN Rel:
    NOT (r IN Rel (nonsense))
END nonsense.
```

The iteration yields

```
{ }
Rel
{ }
```

and has apparently no limit.

However, there are meaningful constructor definitions with negation and universal quantification, and the DBPL compiler will recognize a subclass thereof, defined by the so called positivity constraint.

Let us start with auxiliary definitions:

#### Definition: Names appearing under NOT and ALL

Let  $f$  be a DBPL expression.

A name  $n$  is said to **appear under ALL** if  $f$  is of the form

$f = \dots \text{ALL } r \text{ IN exp } (p(r, \dots)) \dots$

and  $n$  appears in  $\text{exp}$ .

A name  $n$  is said to **appear under NOT** if  $f$  is of the form

$f = \dots \text{NOT fact } \dots$

and  $n$  appears in the factor  $\text{fact}$ .

Note that these definitions may be nested, i.e., a name may appear under several ALLs and NOTs. In

$\text{ALL } r \text{ IN exp } (p(r, \dots))$

a name  $n$  appearing in  $p(r, \dots)$  but **not** in  $\text{exp}$  is **not** considered to appear under **this** ALL.

#### Definition: positivity of a DBPL expression

Let  $f(\text{Rel}_1, \dots, \text{REL}_n)$  be a DBPL expression.

$f$  is said to satisfy the **positivity constraint** if each occurrence of  $\text{Rel}_i$  appears under an **even** total number of negations and universal quantifiers.

The idea of positive expressions is similar to 'safe' expressions in [Ullm 82] by which the definition of infinite relations in relational calculus expressions is avoided.

#### Lemma:

Each DBPL expression  $f(\text{Rel}_1, \dots, \text{Rel}_n)$  that satisfies the positivity constraint is monotonic in all its arguments.

#### Proof Sketch:

Change  $f$  as follows: Replace range-coupled quantifiers by their one-sorted version [JaKo 83]:

$\text{ALL } r \text{ IN Rel } (pred(r, \dots)) = \text{ALL } r (\text{NOT}(r \text{ IN Rel}) \text{ OR } pred(r, \dots))$

$\text{SOME } r \text{ IN Rel } (pred(r, \dots)) = \text{SOME } r (r \text{ IN Rel AND } pred(r, \dots))$

Thus we have replaced each occurrence of  $\text{Rel}_i$  under a universal quantifier by

an occurrence under NOT. Thus, if the number of ALLs plus the number of NOTs over each occurrence of Rel gave an even total, we now have an even number of NOTs over each occurrence Rel of a Rel<sub>i</sub>. However, if this is the case, we can remove the negations, using generalized deMorgan and distribution laws [JaKo 83] to move all NOTs as far into the expression (i.e. to the right) as possible and applying the double negation law NOT(NOT(pred))=pred. The resulting expression will be monotonic in all its arguments.

A similar lemma is given in [ChHa 82]. For simplicity, the DBPL compiler accepts only constructors satisfying the positivity constraint. It should be noted, however, that there are non-monotonic constructors for which the limit of the fixed point computation exists. The following example is derived from [Hehn 84]:

```

TYPE cardrel = RELATION ... OF RECORD number: CARDINAL END;
CONSTRUCTOR strange FOR BaseRel: cardrel (): cardrel;
BEGIN EACH r IN BaseRel:
    NOT SOME s IN BaseRel {strange}
        (r.number=s.number+1)
END strange .

```

Let Rel = {0, 1, 2, 3, 4, 5, 6}. The computation of Rel {strange} through the iteration

```

{}
{0,1,2,3,4,5,6}
{0}
{0,2,3,4,5,6}
{0,2}
{0,2,4,5,6}
{0,2,4}
{0,2,4,6}
{0,2,4,6}
etc.

```

has the limit {0,2,4,6}.

However, examples like this one look very artificial and are much more difficult to understand by programmer and compiler than the simple positivity constraint; they are, therefore, not allowed in DBPL.

### 3.4 Options for Fixpoint Enhancements in Database Programming

In this subsection, we summarize the options for expressing the Least Fixpoint Operator semantics in a database programming language like DBPL. For database programming languages we distinguish six possibilities to include fixpoint operations. Our constructor approach can be seen as the seventh alternative.

- Programm iteration;
- Recursive boolean functions and procedures;
- Specialized LFP operators;
- Equational relation variable declarations;
- Views as relation-valued functions;
- Logic Programming.

The first two options have already been available in early languages such as Pascal/R [Schm 77] although they did not receive much attention there. The

programs for computing the limits in section 3.1 may serve as examples for this approach. Similar effects can also be achieved using recursive functions (to generate recursive relations or to test membership recursively). Both methods share the problem of too much generality since the programmer can write anything into the loop or the function body; this severely limits query optimization.

Query-by-example [Zloo 77] was one of the first systems to contain a **specialized operator** for transitive closure. More recently, the query language QUEL has been augmented with an operator \* which can extend any QUEL command with the semantics "to repeat the command forever" [Kung 84], [IoShWo 84]. [EmEmDo 84] combine a similar approach with view-oriented concepts as described below. While some algebraic optimization of such language extensions is possible [Kung 84], the approach is essentially procedural and does not seem to fit well into a calculus-oriented language.

**Equational relation definition** bears a close similarity to relation definition by constructors. However, instead of constructing relations explicitly from conventionally typed variables, the type concept itself could be extended to allow implicit relation definition by a set of constraining conditions:

```
VAR Infront: infrontrel;
    Ahead: aheadrel ( EACH r IN Infront: TRUE, <f.front,b.back> OF
                    EACH f IN Infront,
                    EACH b IN Ahead: f.back=b.front ) .
```

The work on equational constraint expressions [Morg 84] follows a similar approach.

A number of researchers have proposed parameterized view definitions for query language extensions (e.g., [MaReSc 84], [EmEmDo 84]). From a programming language standpoint, views can be interpreted in two different ways. If relations are considered as generalized tables or arrays, these structures seem to be adequately handled by selectors and constructors. If relations are considered as sets, views can be considered as **relation-valued functions**. Since recursive functions are available in modern programming languages, the extension to relation-valued functions would be small, for example:

```
FUNCTION ahead (Current:aheadrel): aheadrel;
VAR New: aheadrel;
BEGIN
    New := ( EACH r IN Current,
            <c.head,d.tail> OF
            EACH c,d IN Current: c.tail=d.head );
    IF New = Current THEN RETURN Current
    ELSE RETURN ahead(New)
END ahead;
...
Ahead := ahead(Infront).
```

However, as previously discussed, functions are too general to be optimized efficiently. Of course, if used in a pure query language environment such as SQL, relation-valued functions can only define parameterized views and thus may not raise the problems present in tightly integrated database programming languages.

One of the most important areas closely related to our work is that on **logic programming** as exemplified by PROLOG (e.g. [CIme 81]). Being based on Horn clauses, the programming language PROLOG (without cut, fail and negation) can be shown to be equivalent to a data base query language with the least fixed point operator [ChHa 82]. As far as our language extensions are concerned, we

have the following lemma:

**Lemma:**

The constructor mechanism is as powerful as function-free PROLOG without cut, fail, and negation.

**Proof sketch:** Horn clauses are precisely representable by applying a single fixed point operator to a positive existential query [ChHa 82]. Furthermore, mutual recursion can be replaced by a single fixed point operator by moving the mutual recursion into the arguments. Therefore, any query representable in function-free Horn clauses is also representable by the constructor mechanism.

As far as negation is concerned, our approach assumes a closed world and is guaranteed to terminate because of positivity. Therefore, it is not directly comparable with PROLOG's NOT as far as generality is concerned. However, it seems to be more practical because the problem of endless loops is eliminated.

#### 4 Compilation and Optimization of Constructors

In this section, we investigate the implementation of constructors and the optimization of queries in which constructed relations appear. Constructed relations are interpreted as a generalization of the range-nested expressions of [JaKo 83]. First, we study the compilation of queries over constructed relations into queries over base relations; obviously, the most interesting part of this is the handling of recursion. Then, we discuss the optimization of such queries. Rather than adding to the long list of specialized techniques for recursion optimization, we present a three-level framework tailored for the database programming environment in which such techniques can be integrated. For space reasons, details of its implementation must be left to a forthcoming paper.

[JaKo 83] introduced a concept of range nesting for relational calculus expressions. Basically, it allows the substitution of relational expressions for range relations in queries, using the following rules:

- N1: (EACH  $r$  IN  $R$ : pred1 AND pred2)  
     $\Leftrightarrow$   
    (EACH  $r$  IN (EACH  $r'$  IN  $R$ : pred1): pred2)
- N2: SOME  $r$  IN  $R$  (pred1 AND pred2)  
     $\Leftrightarrow$   
    SOME  $r$  IN (EACH  $r'$  IN  $R$ : pred1) (pred2)
- N3: ALL  $r$  IN  $R$  (NOT(pred1) OR pred2)  
     $\Leftrightarrow$   
    ALL  $r$  IN (EACH  $r'$  IN  $R$ : pred1) (pred2)

Selected and constructed relations can be interpreted as methods to **name** such extended range expressions. If we want to follow the  $\Leftrightarrow$  direction in order to understand and optimize a query in terms of base relations, the question becomes by which predicate pred1 to replace the constructed relation. In this subsection, a representation and compilation method for solving this problem are presented. Consider the expression

(EACH  $r$  IN Rel{constr}: pred( $r$ ))

Of course, the easiest solution is to compute Rel{constr} completely by all least fixed points of related constructor definitions and then test pred( $r$ ).

However, propagating the constraints given by  $\text{pred}(r)$  into the constructor definition may considerably reduce query evaluation costs. A case-by-case analysis of various constructor types will demonstrate how this can be done. Assume first that the definition of  $\text{constr}$  does **not** contain any constructed variable, i.e.  $\text{constr}$  works only on base relations.

**Case 1 (Selector):** The constructor definition contains a single relational expression (no union) with a single free variable. In this case, the transformation rules N1 to N3 apply directly, possibly in conjunction with a projection on the target attributes.

**Case 2 (Join):** The constructor definition contains a single relational expression but possibly more than one variable. In this case, substitute  $r.f$  in  $\text{pred}(r)$  by  $x.g$  if  $x.g$  appears in the position  $f$  of the constructor's target list (possibly with renaming).

**Case 3 (Union):** The constructor definition is a union of relational expressions. If  $\text{pred}(r)$  satisfies the positivity constraint, treat each of these relational expressions separately and let the result be the union of the expression values: (If  $\text{pred}(r)$  does not satisfy the positivity constraint, there may be cases where the constructed relation has to be computed fully prior to the evaluation of  $\text{pred}(r)$  [JaKo 83]).

If the tuple variable whose range expression is constructed is existentially or universally quantified, the above rules apply in a similar fashion, corresponding to rules N2 and N3. The rules actually present just a minor generalization of [Ston 75].

Consider now the case that the constructor definition does contain constructed relations. The naive application of the above rules would give an infinite derivation sequence in case of recursive constructors. Adapting a trick described in [Naqv 84], [Venk 84], a finite representation of this infinite sequence can be devised from which appropriate least fixed point computations can be generated. Due to space limitations, we can only sketch the algorithm here, using constructor ahead as an example.

1. Construct an augmented quant graph for the constructors. A quant graph represents a relational calculus query [JaKo 83]; it has a node for each tuple variable with its range definition and a directed arc in quantifier direction (outside in) for each join term and each enforced quantifier sequence. An augmented quant graph is constructed by adding special nodes representing the head of constructors and directed arcs representing the attribute relationships between the result relation and the range definitions as illustrated in Fig. 3.
2. Construct directed arcs from each quantified node with a constructed range relation (in the example:  $b$ ) to the corresponding constructor head; in doing so, check for unifiability of the parameters and the base relation of the constructors. We have now constructed the equivalent of a clause interconnectivity graph [Sick 76].
3. Evaluate each component as follows. For acyclic subgraphs, replace the constructor definitions by subqueries on base relations and optimize as described, e.g., in [Jark 84]. Most cyclic subgraphs correspond to recursion (for exceptions as tautologies see [Sick 76]). We can now either apply the standard algorithms, i.e., LFP computation of the related constructor definitions, recursive calls of iterative procedures [HeNa 84], or a tuple-at-a-time cycling [McSh 81]; or we can attempt to employ capture rules [Ullm 84] to detect special cases such as [Schn 78], [MiNi 83], [Fron 84].



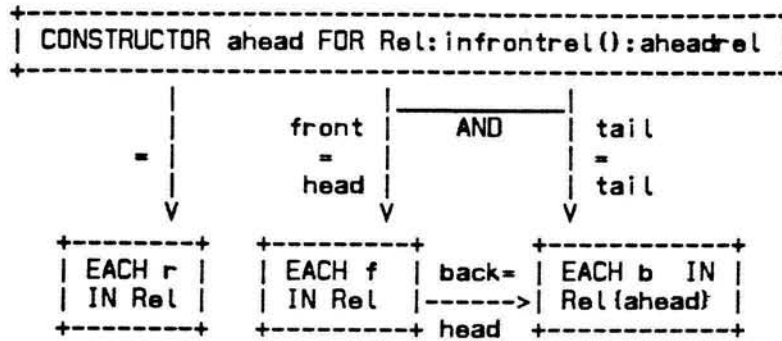


Figure 3: Augmented quant graph

Applying this method at query evaluation time may be quite expensive if many constructed relations are defined. Our optimization strategy tries to move many of these tasks into the compilation phase; this is even more important in a database programming language than in an interactive query language because compilation is usually decoupled from execution.

On the other hand, database programming languages are frequently used to implement higher-level interfaces and therefore contain only incompletely specified query forms rather than full queries. These observations lead to a three-level strategy in the optimization of the system that makes full use of the degrees of information available to different phases of the DBPL compiler and to the runtime support system.

On the **type-checking** level, the compiler performs an analysis of the individual constructor definitions and their relationships. For example, this phase contains the positivity test within the constructor definition. It also constructs a rough version of the extended quant graphs described above. In terms of optimization, one major purpose of this is to offer a preliminary partitioning of the set of constructor definitions in disconnected graphs.

This partitioning can be done by stepwise refinement. A first version of the graph would just mention relation and constructor names. If some of the remaining partitions are still very large, they could then be refined to an intermediate level that, e.g., distinguishes between free and bound variables [Ullm 84].

On the **query compilation** level, the compiler looks at the query forms appearing in the database program. These query forms may use range relations that apply constructors to base relations, selected relations, or constructed relations. The compiler can now instantiate the appropriate constructor definition graphs and complete the construction of full extended quant graphs for each query. If such a graph contains a recursive cycle, the compiler can generate an appropriate version of the fixed point algorithm [HeNa 84], [Ullm 84]. For non-recursive queries, full decompilation and view optimization are performed.

The discussion overlooks the fact that constructor and selector definitions may contain parameters. In case these are constant values in restrictive terms of constructor definition or associated query, we can represent this situation by defining an appropriate selector. This selector will provide a logical or even physical access path for instantiations of the parameters. A logical access path is a compiled procedure with dummy constants [HeNa 84]. A physical access

path actually materializes a relation corresponding to the query with the constants used as variables, and partitions it according to the different constant values. Obviously, a physical access path would be generated only in case of heavy query usage since unrestricted constructed relations may be very large. Maintenance for such access paths is discussed in [ShTZ 84].

If the parameters are of type relation, they may be instantiated at runtime with constructed relations, possibly leading to the merging of previously independent subgraphs. At a rather high cost, the compiler can check whether suitable constructors have actually been defined in the database program. In any case, this case will only permit partial logical access paths to be generated at compilation time.

Finally, the **runtime support** subsystem of query processing must help in the evaluation of fully instantiated queries. In some cases, this will just mean the execution of the compiled database programs. In the case of selectors generated at compile time, it means the generation and utilization of physical access paths. In the case of relation parameters, it may mean the integration of pieces of precompiled definitions into meaningful database programs. A major advantage of the DBPL environment over, e.g., a PROLOG environment is that all of these tasks can be formulated elegantly with the existing language tools and are executed in a set-oriented constructive fashion rather than by tuple-oriented theorem proving.

## 5 Conclusion

Relational database systems are based on first order logic and provide, within that framework, solutions for many technical problems with data-intensive applications, such as query optimization, concurrency management, and data distribution. While AI-oriented systems have their main emphasis on issues of knowledge representation and reasoning, future applications will require the same technical support for problems originating from large scale fact and rule management.

We argue that the cutting point should make the DBMS responsible for as much efficient mass-processing of data as possible, whereas the AI system retains the responsibility for the more subtle tasks, such as handling open worlds (i.e., incomplete knowledge and non-monotonic reasoning) for which intelligent and frequently problem-specific heuristics are needed since the problem in general is computationally intractable or even undecidable [BrLe 84].

In this paper, we propose an extension of the relational approach that can handle nested and recursive rule definition and evaluation adequately and efficiently. In an orthogonal approach to data model extension we investigate object structures that allow nested and recursive structure definition and component selection [Lame 84], [LaMuSc 84]. Properly integrated, we expect from both kinds of research a new generation of data models for object-oriented, rule intensive applications.

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