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<u>ABSTRACT</u>. The traditional Query Modification approach to query processing is inappropriate for views involving <u>generalization</u>. We use a combination of modification and <u>materialization</u> for queries over such views. Furthermore, by choosing modification or materialization as part of global optimization, we permit more optimization than would be provided by a purely modifying approach.

#### INTRODUCTION

<u>Generalization</u> is an abstraction that groups conceptually related objects into a "generic" object [SMITH]. For example, students and instructors can be generalized as people.

We can devise views that represent generalization [KATZ], [DAYAL\_1]. However, as will be seen, the standard approach of Query Modification [STONEBRAKER] will not correctly handle all queries over such views. Furthermore, even where it works, query modification may not be the best way to handle such queries.

We have developed a query processing architecture that correctly and efficiently handles these queries.

Although we developed this approach in terms of the Functional Data Model and the language DAPLEX [SHIPMAN], our observations and algorithm apply to any system supporting virtual generalization.

### A SIMPLE DATA MODEL AND LANGUAGE

Consider a database of entities. Each entity

is of a <u>type</u> that specifies the name and range of its <u>attributes</u>. An attribute value can be a string, integer, a reference to another entity (perhaps of a different type) or a set of any of these.

Figure 1 is a schema for such a database.

Using the notation of [SHIPMAN], queries over such a database can be expressed, procedurally, with statements that <u>create</u> new entities, <u>print</u> attribute values and <u>iterate</u> through sets of entities (perhaps in a specified order).

For example, the following query prints the name of each cat followed by the name (alphabetically) of each of its "friends":

```
for each A in ANIMAL
where ANIMAL_SPECIES (A) = "Cat"
loop
print (ANIMAL_NAME (A));
for each F in ANIMAL
where F is in ANIMAL_FRIENDS (A)
ordered by ANIMAL_NAME (F)
loop
print (ANIMAL_NAME (F));
end loop;
end loop;
```

Expressions can contain <u>aggregations</u>. These are built-in operations over a set of values. A query to print the median INSTRUCTOR\_SALARY taken over the set of INSTRUCTORs would be: "print (median (INSTRUCTOR\_SALARY (INSTRUCTOR)))".

### VIEWS WITH GENERALIZED ENTITIES

Our system supports two kinds of virtual entities. <u>Simple</u> types are derived from <u>one</u> underlying entity type. For each underlying entity (subject to selection), the attributes of that entity are used to create one of the virtual entities.

For example, as shown in Figure 2, CAT entities can be derived from ANIMALS.

We also support <u>generalized</u> types. These represent the "generalization" of <u>several</u> underlying types of entities. For example, the type "PERSON" can generalize STUDENTS and INSTRUC-TORs.

As described in [DAYAL\_1] and [GOLDHIRSCH],

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each element of the <u>bi-directional outer-join</u> [CODD], [DAYAL\_2] of the STUDENTS and INSTRUC-TORs can be used to create a PERSON entity (assuming we know when a STUDENT and INSTRUCTOR represent the same "person").

A view specifying PERSONs and DWELLINGs is shown in Figure 3. (The derivation for DWELLING is similar to that of PERSON and is omitted.) In general, an entity X generalizing <u>u</u> underlying

types X1,...,Xu is defined with 2<sup>u</sup>-1 derivation <u>components</u> (corresponding to the non-empty subsets of {X1,...,Xu}).

# MODIFICATION DOES NOT ALWAYS WORK

There are two ways to handle a query's reference to virtual entities: <u>materialize</u> the entities and then run the query; or, syntactically <u>modify</u> the query to refer to the underlying (non-virtual) entities.

Materialization always works. Since derivations are syntactically similar to queries, an unimaginative system can simply "execute" the derivation before running the query (we do NOT recommend this approach).

The <u>modification</u> approach was introduced, in a relational context, without generalizations, in [STONEBRAKER]. Figure 4b shows the result of using modification to eliminate the CAT loop of Figure 4a.

Observe that the effect of the "ordered by" clause was preserved through the modification.

[KATZ] and [DAYAL\_1] propose a modification approach to handle queries over <u>generalizations</u>. The idea is that, for each iteration over a generalization (of <u>u</u> underlying types), the query

is broken up into 2<sup>u</sup>-l separate queries -- one for each <u>component</u> of the derivation. If the resulting queries refer to other generalizations, the algorithm is applied recursively.

Figure 5a contains a query that prints (without ordering) the EARNINGS of each PERSON. After modification, using the derivation of Figure 3, we get the three queries of Figure 5b.

The modification approach of [KATZ] and [DAY-AL\_1] does NOT correctly handle queries containing <u>ordered iterations</u> and/or <u>aggregations</u> over generalizations.

To see why ordering isn't correctly modified, imagine adding an "ordered by PERSON\_NAME (P)" clause to the loop in Figure 5a. The clause would appear, referring to STUDENT and/or IN-STRUCTOR, in EACH of the queries of Figure 5b. This produces three independently ordered lists, rather than the single, ordered list printed by the query in Figure 5a.

To see why aggregation fails, observe that there is no way to "decompose" a median over the three <u>components</u> of PERSONS. That is, no combination of medians over STUDENTs and INSTRUCTORs can be used to modify the query "print (median (PERSON\_EARNINGS (PERSON)))".

(Some aggregations CAN be handled by modification. For example, the <u>count</u> of the PERSONs is the <u>sum</u> of: the <u>count</u> of PERSONs derived

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from STUDENTs, the <u>count</u> of PERSONs derived from INSTRUCTORS and the <u>count</u> of PERSONS derived from both a STUDENT and an INSTRUCTOR.)

# MODIFICATION IS NOT ALWAYS DESIRABLE

The "obvious" solution might be to materialize generalizations that are being aggregated and/or ordered, and to use modification for all other virtual loops.

An assumption implicit in [STONEBRAKER], [KATZ] and [DAYAL\_1] is that, given a choice, <u>modification</u> is the preferred method. This is true for SIMPLE types.

However, because it drastically alters a query's architecture, modification for generalizations can produce inefficient queries.

As an example, consider the query in Figure 6a. Observe that the loops over PERSON and DWELLING can each be handled either by modification or materialization.

An optimization goal might be to minimize the number of joins [RIES] needed to process the nested loops in this query. In this example, we can show that it takes more joins to run the modified queries than it would to materialize PERSON and DWELLING (running the query as is).

Modifying the loop over PERSON results in 3 queries each over DWELLING, the components of PERSON (STUDENT and INSTRUCTOR) and CAT. Each query has 4 loops and represents 3 joins.

For each query, applying modification to eliminate the DWELLING loop produces 3 queries each over STUDENT, INSTRUCTOR, CAT and the components of DWELLING. This is a total of 3\*3=9 queries. Each query contains 5 loops and represents 4 joins. The total number of joins is 9\*4=36.

Assume we use only <u>modification</u> for a query of <u>L</u> loops. Assume that <u>g</u> of these loops range over generalizations of <u>u</u> underlying types. The

result is  $(2^{u}-1)^{g}$  queries each with (L + g(u-1)) loops. This represents a total of

$$(2^{u} - 1)^{g}$$
 (L + g(u-1) - 1) =  
0 (2<sup>gu</sup>) joins.

On the other hand, using a "worst case" strategy (see [GOLDHIRSCH]), we can materialize the PERSONs and DWELLINGs each using 6 joins. The query itself contains 3 loops and represents 2 joins. So, to materialize PERSON, materialize DWELLING and then run the original query over the materialized entities, we need (6+t+2) = -14 joins.

In general, assuming a worst case strategy, we can handle a query of <u>L</u> loops (of which <u>g</u> ar) generalizations of <u>u</u> underlying types) using.

 $g(2^{u} - 1)(u-1) + L - L =$ 

 $0 (g2^{u})$  joins.

As the number of generalizations in a query

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grows, the number of joins after modification grows <u>exponentially</u>. The number of joins for <u>materialization</u> grows only <u>linearly</u>.

The point is not that materialization is therefore "better" than modification, or visaversa, but that there are cases where each might be better than the other. Arbitrarily preferring one method can significantly detract from the optimization of the query.

# A QUERY PROCESSING ARCHITECTURE

We need a processing scheme that sensibly chooses <u>modification</u> or <u>materialization</u> for each loop (over a generalization) in a query.

The traditional plan works in three steps: (1) <u>view-map</u> the query into one or more queries that refer only to existing entities, (2) <u>op-</u> <u>timize</u>, or produce a "strategy" for running the resulting query (queries) and (3) <u>execute</u> the query (queries) according to the strategy.

This architecture is inappropriate since (a) the view-mapper cannot understand the impact of changing the architecture of the query -- nor can it calculate the cost of any particular materialization, and (b) an ordinary optimizer would not know the view-mapping algorithm.

We choose to make step (2) a "hybrid" of optimization and <u>deferred modification</u>. As an illustration, consider the processing of the query in Figure 6a.

The <u>view-mapper</u> proceeds, iteration by iteration, to resolve references to virtual entities. Say, for example, it first recognizes CAT as a SIMPLE virtual type. It <u>modifies</u> the CAT loop, creating an appropriately restricted loop over ANIMAL.

Figure 6b shows the query after the modification of the CAT loop.

Notice that "A is in PERSON\_CATS (P)" is, at this stage of the view-mapping, typeinconsistent: the range of PERSON\_CATS is the set of CATs, not the set of ANIMALS.

The view-mapper now sees the loops over generalizations DWELLING and PERSON. With each of these, it associates a <u>copy</u> of the corresponding BOJ+ derivation.

To avoid the type inconsistencies, we <u>view</u>map these copies. In particular, the definition of PERSON\_CAT (in PERSON's BOJ+) refers to virtual type CAT. Since CAT is a SIMPLE type, the definition is handled by <u>modification</u>. For example, where a PERSON is both a STUDENT and a INSTRUCTOR, the definition becomes:

```
PERSON_CATS :=
    {A in ANIMAL where ANIMAL_SPECIES(A) = "Cat"
        and
        (ANIMAL_NAME(A) is in
        STUDENT_ANIMAL_NAMES(S)
        or
```

ANIMAL\_NAME(A) is in INSTRUCTOR\_ANIMAL\_NAMES(I))}

After the view-mapping of PERSON's derivation, the range of PERSON\_CATS has effectively

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been changed from CATs to ANIMALs. The viewmapped query is now type-consistent.

The <u>query optimizer</u> now considers all (or many) of the different processing strategies for the query. This includes choosing <u>materializa-</u> <u>tion</u> or <u>modification</u> for each loop over a generalization.

To make this choice, the optimizer first decides which loops <u>must</u> be handled by materialization (none, in this example). For each remaining loop, it uses the Deferred Modifier to compare the (global) effects of using modification with the cost of materializing the generalization.

In this example, if nested loops are processed by joins, the optimizer might decide to materialize both PERSONs and DWELLINGS.

### CONCLUSIONS

The ability to define and query virtual generalizations is desirable. Our approach for handling queries over virtual entities is significantly different from previous proposals:

1. To handle generalized entities <u>correctly</u>, we use both materialization and modification.

2. To do so <u>efficiently</u>, our optimizer collaborates with the view-mapper to decide whether an entity is to be materialized or modified.

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string;
integer;
set of strings;
-
string;
integer;
set of strings;
string;
string;
set of ANIMALs;

# Figure 1. A Database Schema.

string;

CAT

cat\_name:

derive CAT from
for each A in ANIMAL
where ANIMAL\_SPECIES = "Cat"
loop
create CAT

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(CAT NAME := ANIMAL NAME (A)); end loop; end derive; Figure 2. A View with a SIMPLE derivation. PERSON person\_name: string; person\_earnings: integer; person\_cats: set of CATs; DWELLING dwelling\_occupants: set of strings; dwelling\_addr: string; derive PERSON from for P in BOJ+ (S in STUDENT, I in INSTRUCTOR) using join predicate: "STUDENT NAME (S) = INSTRUCTOR NAME (I)" case P is a STUDENT but not an INSTRUCTOR: PERSON NAME := STUDENT NAME (S). PERSON EARNINGS := - STUDENT EXPENSES (S), PERSON CATS := {C in CAT where CAT NAME (C) is in STUDENT\_ANIMAL\_NAMES (S)} case P is an INSTRUCTOR but not a STUDENT: PERSON NAME := INSTRUCTOR NAME (1), PERSON\_EARNINGS:= INSTRUCTOR\_SALARY (1), PERSON CATS:= {C in CAT where CAT NAME (C) is in INSTRUCTOR ANIMAL NAMES (1)} case P is both a STUDENT and an INSTRUCTOR: PERSON NAME := STUDENT NAME (S), PERSON EARNINGS := INSTRUCTOR SALARY (I) - STUDENT EXPENSES (S), PERSON CATS:= {C in CAT where CAT NAME (C) is in INSTRUCTOR ANIMAL NAMES (1) or CAT\_NAME (C) is in STUDENT ANIMAL NAMES (S) end loop; end derive; Figure 3. A View with GENERALIZED derivations. for each C in CAT ordered by CAT\_NAME (C) 1000 print (CAT\_NAME (C)); end loop; Figure 4a. A Query over SIMPLE type CAT. for each A in ANIMAL where ANIMAL\_SPECIES (A) = "Cat" ordered by ANIMAL NAME (A) 1000 print (ANIMAL NAME (A)); end loop; Figure 4b. After the query has been MODIFIED. Proceedings of the Tenth International

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for each P in PERSON loop print (PERSON EARNINGS (P)); end loop; Figure 5a. A Query over GENERAL type PERSON. for each S in STUDENT where for every I in INSTRUCTOR: INSTRUCTOR\_NAME (I)  $\neq$  STUDENT\_NAME (S) 1000 print (- STUDENT EXPENSES (S)); end loop; for each I in INSTRUCTOR where for every S in STUDENT: STUDENT\_NAME (S)  $\neq$  INSTRUCTOR\_NAME (I) 1000 print (INSTRUCTOR SALARY (1)); end loop; for each S in STUDENT loop for each I in INSTRUCTOR where INSTRUCTOR\_NAME (I) = STUDENT\_NAME (S) 100p print (INSTRUCTOR SALARY (I) - STUDENT\_EXPENSES (S)); end loop; end loop; Figure 5b. The same query, after MODIFICATION for each D in DWELLING loop for each P in PERSON where PERSON\_NAME (P) is in DWELLING\_OCCUPANTS (D) 100p print (PERSON\_NAME (P)); for each C in CAT where C is in PERSON\_CATS (P) ordered by CAT\_NAME (C) 1000 print (CAT\_NAME (C)); end loop; end loop; end loop; Figure 6a. A Query over several entity types for each D in DWELLING loop for each P in PERSON where PERSON\_NAME (P) is in DWELLING\_OCCUPANTS (D) 1000 print (PERSON\_NAME (P)); for each A in ANIMAL where ANIMAL\_SPECIES (A) = "CAT" and A is in PERSON\_CATS (P) ordered by ANIMAL\_NAME (A) 1000 print (ANIMAL NAME (A)); end loop; end loop; end loop;

Figure 6b. The query after MODIFYING for CATs.

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