

Theoretical Aspects of Schema Merging

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Abstract

A general technique for merging database schemas is developed that has a number of advantages over existing techniques, the most important of which is that schemas are placed in a partial order that has bounded joins. This means that the merging operation, when it succeeds, is both associative and commutative, *i.e.*, that the merge of schemas is independent of the order in which they are considered — a property not possessed by existing methods. The technique is interactive in that users make assertions about the relationships between the nodes of the schemas to be merged. These assertions are then considered to be elementary schemas, and are combined with the schemas using precisely the same merging operation.

The technique is general and can be applied to a variety of data models. It can also deal with certain cardinality constraints that arise through the imposition of keys. A prototype implementation, together with a graphical interface, has been developed.

Keywords: schemas, merging, semantic data models, entity-relationship data models, inheritance

1 Introduction

The desire to provide user views that combine existing databases, and to combine user views during the design process of new databases, leads directly to the problem of schema merging — a problem that has been present in the database literature for at least ten years and to which a variety of sophisticated techniques have been applied. At one end of the spectrum, the user is provided with a set of tools for manipulating two schemas into some

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form of consistency [1, 2]; at the other end, algorithms have been developed that take two schemas, together with some constraints, and create a merged schema [3]. In general, one will want to use a method that lies somewhere between these two extremes; a number of such variations have been explored, and are surveyed in [4]. It appears that some user manipulation of the given schemas is essential — especially to introduce consistent names — but that a merging algorithm can also be very useful, especially when large schemas are involved.

To the best of our knowledge, the question of what meaning or semantics this merging process should have has not been explored. Indeed, several of the techniques that have been developed are *heuristics*: there is no independent characterization of what result they should produce. One would like to have some semantic basis for a merge that would characterize it with some notion of consistency with the associated data. In this paper, we shall develop a simple and general characterization of database schemas that allows us to give natural definitions of what a merge is in terms of the informational content of the schemas being merged. In particular we shall define a kind of merge which takes the union of all the information stored in a collection of database schemas, and, wherever possible, forms a schema presenting this but no additional information. We shall be able to define a binary merging operator that is both commutative and associative, which means that if two or more schemas are merged, the result is independent of the order in which the merges are performed. Existing techniques do not have this property. Worse still, lack of associativity is endemic to the data models against which the merging process is commonly defined (for example, the Entity-Relationship (ER) model).

Using a more general formalism, we will be able to rescue this situation by introducing special, additional information during the merging process. The additional information describes its own origin, and can be readily identified to allow subsequent merges to take place. In addition, our new schema merging technique may be applied to other, existing models, such as the ER-model, by first translating the schemas of these models into our model, then carrying out the merge, and finally translating back into the original model. It is possible to show that, if such an approach is used, then the merging process respects the original model.

The paper is organized as follows: we shall first describe a general data model that subsumes, in some sense, data models such as relational, entity-relationship and functional. We then observe that even in this general model, we cannot expect to have associative merges. We relax the constraints on our model to produce a still more general formulation of a *weak* schema for which the merging process is well behaved, and then show how to convert these weak schemas back into our original schemas. We also show that certain common forms of constraints on the schemas, such as key constraints and some cardinality constraints, can be handled in the same framework. Finally we shall indicate how these methods could be used equally well to give definitions of alternative kinds of merge, and, in particular, describe how we could form *lower merges* representing the intersection of the information represented by a collection of schemas.

2 The Model

We represent a schema as a directed graph, subject to certain restrictions, whose nodes are taken from a set \mathcal{C} of *classes*, and with two kinds of edges that are used to represent “attribute of” relationships or “specialization of” relationships between classes. Attribute edges have labels taken from a set \mathcal{L} , so that we represent the attribute-of edges by a relation $\mathcal{E} \subseteq \mathcal{C} \times \mathcal{L} \times \mathcal{C}$. If $(p, a, q) \in \mathcal{E}$ then we write $p \xrightarrow{a} q$, with the meaning that any instance of the class p must have an a -attribute which is a member of the class q . Since in some data models, like the ER data model, the term “attribute” is used to designate a certain kind of node in a schema, we shall use the neutral term *arrow* to refer to these labeled relationships, and say for $p \xrightarrow{a} q$ that p has an a -arrow of class q . The specialization edges are represented by a relation \mathcal{S} on classes; we use the notation $p \Longrightarrow q$ and say that p is a *specialization* of q when $(p, q) \in \mathcal{S}$. This indicates that all the instances of p are also instances of q . Formally, a schema over \mathcal{C}, \mathcal{L} is a triple $(\mathcal{C}, \mathcal{E}, \mathcal{S})$ where $\mathcal{C} \subseteq \mathcal{C}$ is a finite set of classes, \mathcal{S} is a partial order (a reflexive transitive and antisymmetric relation on \mathcal{C}), and \mathcal{E} is a subset of $\mathcal{C} \times \mathcal{L} \times \mathcal{C}$ satisfying

1. If $p \xrightarrow{a} q_1$ and $p \xrightarrow{a} q_2$ then $\exists s \in \mathcal{C} . s \Longrightarrow q_1$ and $s \Longrightarrow q_2$ and $p \xrightarrow{a} s$.
2. If $p \Longrightarrow q$ and $q \xrightarrow{a} r$ then $p \xrightarrow{a} r$.
3. If $p \xrightarrow{a} s$ and $s \Longrightarrow r$ then $p \xrightarrow{a} r$.

for all $a \in \mathcal{L}$ and $p, q, r, s \in \mathcal{C}$

The first constraint says that for any a -arrow of a class p , there is a *least* class s (under the ordering \mathcal{S}) such that a has class s . Such a class is said to be the **canonical** class of the a -arrow of p . The second constraint says that, if q has an a -arrow of class r and p is a specialization of q , then p must also have an a -arrow with class r . The third constraint says that, if p has an a -arrow of class s and s is a specialization of r , then p also has an a -arrow of class r , so constraints 2 and 3 together mean that arrows are, in some sense, preserved by specialization. It is worth remarking that we could equally well have defined the arrows as partial functions from classes to classes, which is how they are expressed in Dayal’s definition of a functional schema [2]. If we write $p \xrightarrow{a} q$ when p has an a -arrow with canonical class q , we have the conditions

- D1.** If $p \xrightarrow{a} q_1$ and $p \xrightarrow{a} q_2$ then $q_1 = q_2$
- D2.** If $q \xrightarrow{a} s$ and $p \Longrightarrow q$ then $\exists r \in \mathcal{C} . r \Longrightarrow s$ and $p \xrightarrow{a} r$

Further, given any \longrightarrow satisfying conditions D1 and D2, if we define the relation \longrightarrow by $p \xrightarrow{a} q$ iff there exists a $s \in \mathcal{C}$ such that $s \Longrightarrow q$ and $p \xrightarrow{a} s$, then \longrightarrow will satisfy conditions 1, 2 and 3 above. Conditions D1 and D2 are those given for the arrow in [2]

and are also given by Motro [1] as axioms for functional schemas (the latter uses unlabeled arrows).

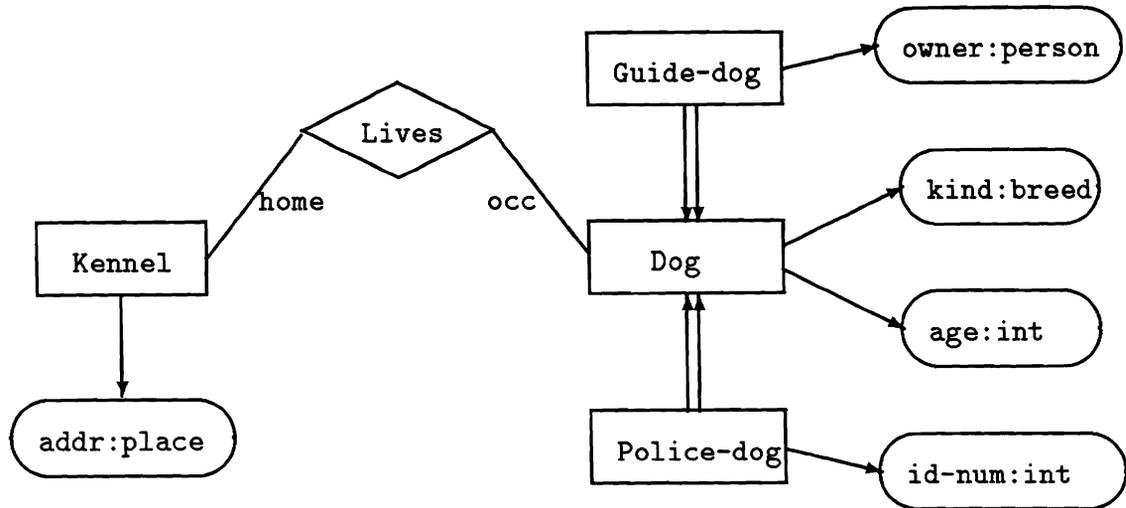


Figure 1: An ER-diagram with “isa” relations

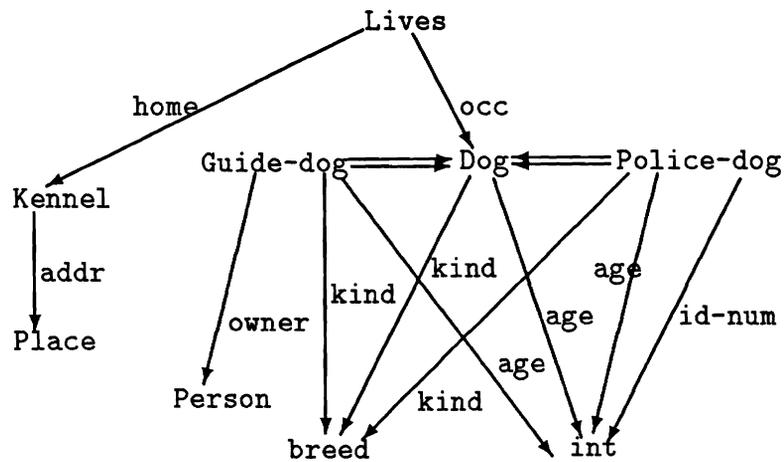


Figure 2: A database schema with “isa” relations

For example, the ER diagram shown in figure 1 corresponds to the database schema shown in figure 2, where single arrows are used to indicate edges in \mathcal{E} and double arrows are used to represent pairs in \mathcal{S} (double arrows implied by the transitivity and reflexivity of \mathcal{S} are

omitted). In all the subsequent diagrams, edges in \mathcal{E} implied by constraint 2 above, will also be omitted.

Suitable restrictions of such graphs may be used to describe instances of a variety of data models: relational, entity-relationship and functional. For a relational instance, we stratify \mathcal{C} into two classes \mathcal{C}_R and \mathcal{C}_A (relations and attribute domains), disallow specialization edges, and restrict arrows to run labeled with the name of the attribute from \mathcal{C}_R to \mathcal{C}_A (first normal form), while, for the E-R model, we stratify \mathcal{C} into three classes (attribute domains, entities and relationships) and again place certain restrictions on the edges. Moreover, it can be shown that the merging process described in section 4 preserves these restrictions, so that we can merge schemas from other models by first translating them into our model, then merging them, and finally translating them back into the original data model (see [5] for details). By a less constrained process we can describe instances of the functional model [6, 2, 1]. The graphs are also general enough to represent databases with higher order relations (that is, relationships between relationships), and complex data structures (such as circular definitions of entities and relationships), features that are commonly found in object-oriented data models. Consequently, despite its apparent simplicity, the generality of the model makes it a good candidate for schema merging. One should note, however, that further adornment of these graphs is needed to describe instances of sophisticated data models such as those proposed in [7] and [8], which contain constructors for sets and variants.

3 Problems with finding merges of schemas

The first problem to be resolved when forming a common merged schema for a number of distinct databases is to state the correspondences between the classes and correspondences between the arrow labels of the various databases. This problem is inherently *ad hoc* in nature, and depends on the real-world interpretations of the underlying databases. Therefore, the designer of the system must be called on to resolve naming conflicts, whether homonyms or synonyms, by renaming classes and arrows where appropriate. The interpretation that the merging process places on names is that if two classes in different schemas have the same name, then they are the same class, regardless of the fact that they may have different arrow edges. For example, if one schema has a class **Dog** with arrow edges **License#**, **Owner** and **Breed**, and another schema has a class **Dog** with arrow edges **Name**, **Age** and **Breed**, then the merging process will collapse them into one class with name **Dog** and arrow edges **License#**, **Owner**, **Name**, **Age**, and **Breed**. It is also possible to constrain the merging process by introducing specialization relations $a_1 \implies a_2$ between nodes a_1 in schema S_1 and a_2 in schema S_2 . We can treat $a_1 \implies a_2$ as an atomic schema that is to be merged with S_1 and then with S_2 . Because our schema merge is associative and commutative, the result is well-defined; indeed an arbitrary set of constraints can be added in this fashion.

For the remainder of this section and the following section, we will consider the merge of a

collection of schemas to be a schema that presents all the information of the schemas being merged, but no additional information (although in Section 6, we will indicate that there may be other, equally valid interpretations of what the merge should be). Hence what we will consider to be the merge is the “least upper bound” of the database schemas under some sort of information ordering. Recall that, in addition to defining a view of a database, a database schema expresses certain requirements on the structure of the information stored in the database. When we say that one database schema presents more information than another, we mean that any instance of the first schema could be considered to be an instance of the second one. The first schema must, therefore, dictate that any database instances must contain at least all the information necessary in order to be an instance of the second schema. It is clear that, were we to construct an ordering on schemas which represented the fact that one schema contained more information than another, it should be a partial ordering on schemas, and the binary merge of two schemas would be their least upper bound, or join, under this ordering. This ties in well with our intuition that a binary merging operator on schemas should be associative, commutative and idempotent: indeed any such binary operator will lead us to a partial ordering on schemas.

In looking for a suitable information ordering and definition of merge for schemas, we soon find that there are some problems. One of the first problems we notice is that the merge of two schemas may contain extra *implicit* classes in addition to the classes of the schemas being merged. For example, figure 3 shows two schemas being merged. The first schema asserts that the class C is a subclass of both the classes A1 and A2. The second schema asserts that the classes A1 and A2 both have a-arrows, of classes B1 and B2 respectively. Combining this information, as we must when forming the merge, we conclude that C must also have an a-arrow, and that this arrow must be of both the class B1 and B2. Consequently, due to the restrictions in our definition of database schemas in Section 2, the a-arrow of the class C must have a class which is a specialization of both B1 and B2 and so we must introduce such a class into our merged schema.

When we consider these implicit classes further we find that it is not sufficient merely to introduce extra classes into a schema with arbitrary names: the implicit classes must be treated differently from normal classes. Firstly, if we were to give them the same status as ordinary classes we would find that binary merges are not associative. For example consider the three simple schemas shown in figure 4. If we were to first merge the schemas G1 and G2 we would need to introduce a new implicit class (X?) below D and E, and then merging with G3 would make us introduce another new class below X? and F, yielding the first schema shown in figure 5. On the other hand, if we were to merge G1 with G3 and then merge the result with G2, we would first introduce an implicit class below E and F, and then introduce another one below this one and D. Clearly what we really want is *one* implicit class which is a specialization of all three of D, E and F.

Another problem is that it is possible for one schema to present more information than another without containing as many implicit classes. It is clear that for one schema to present all the information of another (plus additional information) it must have, at least,

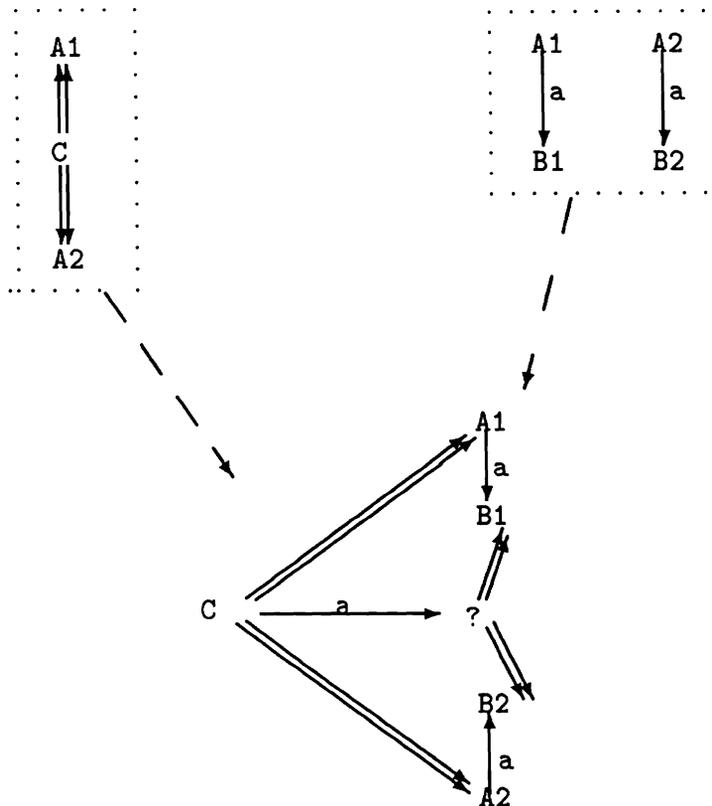


Figure 3: Schema merging involving implicit classes

all the normal classes of the other. However let us consider the two schemas shown in figure 6. We would like to assert that the schema **G3** shown in figure 7 is the merge of the two schemas, **G1** and **G2**, but the schema **G4** also presents all the information of **G1** and **G2**, and in addition contains fewer classes than **G3**. The point is that **G4** asserts that the *a*-arrow of **F** has class **E**, which may have restrictions on it in addition to those which state that it's a subclass of both **C** and **D**, while **G3** only states that the *a*-arrow of **F** has both classes **C** and **D**.

4 Merging Database Schemas

In order to avoid the complexities of introducing implicit classes, we will weaken our definition of database schemas so that implicit classes become unnecessary. We then define an information ordering on these *weak schemas*, such that binary joins do exist and are associative, and form the weak schema merge. Finally we convert the merged weak schema

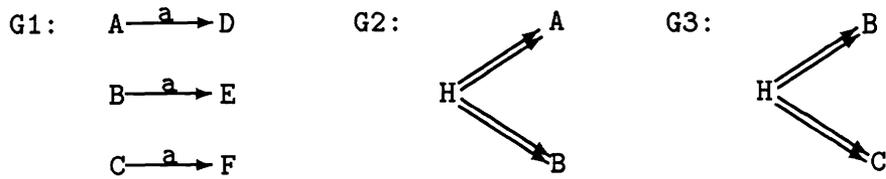


Figure 4: Some simple schemas

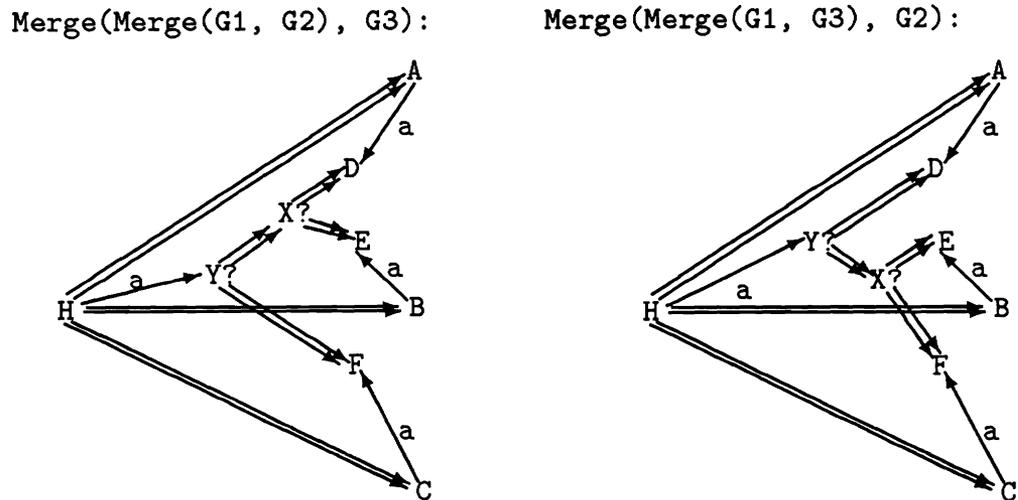


Figure 5: An example of non-associative merging

into a proper schema by introducing additional implicit classes (we will refer to the schemas satisfying the conditions in section 2 as *proper schemas*). The idea is that we can do all our merging using weak schemas, and then convert the result to a proper schema when we are done.

4.1 Weak Schemas

A **weak schema** is a schema in which we no longer require that if class p has an a -arrow, then the a -arrow has a canonical class (condition 1 of proper schemas). Formally, a weak schema over \mathcal{C} , \mathcal{L} is a triple $(C, \mathcal{E}, \mathcal{S})$ where $C \subseteq \mathcal{C}$ is a set of classes, \mathcal{S} is a partial order (a reflexive transitive and antisymmetric relation on C), and \mathcal{E} is a subset of $C \times \mathcal{L} \times C$ satisfying

W1. If $p \Rightarrow q$ and $q \xrightarrow{a} r$ then $p \xrightarrow{a} r$.

W2. If $p \xrightarrow{a} s$ and $s \Rightarrow r$ then $p \xrightarrow{a} r$.

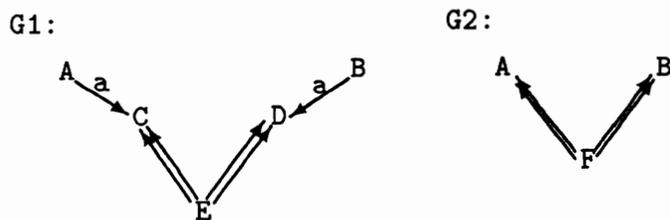


Figure 6: Yet more schemas

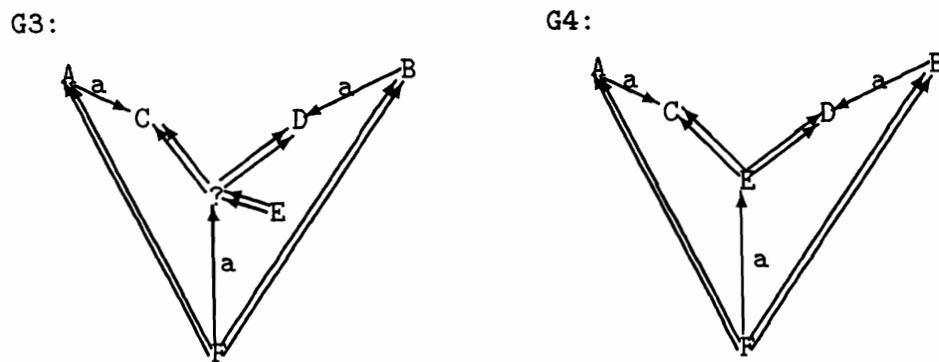


Figure 7: Possible candidates for the merges of the schemas

for all $a \in \mathcal{L}$ and $p, q, r, s \in C$

The ordering on weak schemas is defined in the obvious way: Given two weak schemas $\mathcal{G}_1 = (C_1, \mathcal{E}_1, \mathcal{S}_1)$ and $\mathcal{G}_2 = (C_2, \mathcal{E}_2, \mathcal{S}_2)$, we write $\mathcal{G}_1 \sqsubseteq \mathcal{G}_2$ iff

1. $C_1 \subseteq C_2$
2. $\mathcal{E}_1 \subseteq \mathcal{E}_2$
3. $\mathcal{S}_1 \subseteq \mathcal{S}_2$

That is, every class in \mathcal{G}_1 appears in \mathcal{G}_2 , every a -arrow edge in \mathcal{G}_1 appears in \mathcal{G}_2 , and every specialization edge in \mathcal{G}_1 appears in \mathcal{G}_2 .

It is clear that \sqsubseteq is a partial ordering on weak schemas; it is also bounded complete, as shown in the following proposition.

Proposition 4.1 *For any weak schemas \mathcal{G}_1 and \mathcal{G}_2 , if there exists a weak schema \mathcal{G}' such that $\mathcal{G}_1 \sqsubseteq \mathcal{G}'$ and $\mathcal{G}_2 \sqsubseteq \mathcal{G}'$ then there is a least such weak schema $\mathcal{G}_1 \sqcup \mathcal{G}_2$.*

Proof: Given weak schemas \mathcal{G}_1 and \mathcal{G}_2 as above, define $\mathcal{G} = (C, \mathcal{E}, \mathcal{S})$ by

$$\begin{aligned} C &= C_1 \cup C_2 \\ \mathcal{S} &= (\mathcal{S}_1 \cup \mathcal{S}_2)^* \\ \mathcal{E} &= \{p \xrightarrow{a} s \in (C \times \mathcal{L} \times C) \mid \exists q, r \in C \cdot p \Rightarrow q \in \mathcal{S}, r \Rightarrow s \in \mathcal{S}, \\ &\quad \text{and } q \xrightarrow{a} r \in (\mathcal{E}_1 \cup \mathcal{E}_2)\} \end{aligned}$$

(where $(\mathcal{S}_1 \cup \mathcal{S}_2)^*$ denotes the transitive closure of $(\mathcal{S}_1 \cup \mathcal{S}_2)$, and \mathcal{E} adds edges to $\mathcal{E}_1 \cup \mathcal{E}_2$ necessary for conditions W1 and W2 to hold). It is clear that, if \mathcal{G} is a weak schema, then it is the least weak schema such that $\mathcal{G}_1 \sqsubseteq \mathcal{G}$ and $\mathcal{G}_2 \sqsubseteq \mathcal{G}$. Hence it is sufficient to show that, if there is a weak schema, $\mathcal{G}' = (C', \mathcal{E}', \mathcal{S}')$, such that $\mathcal{G}_1 \sqsubseteq \mathcal{G}'$ and $\mathcal{G}_2 \sqsubseteq \mathcal{G}'$, then \mathcal{G} is indeed a weak schema. The only way that \mathcal{G} can fail to be a weak schema is if the relation \mathcal{S} fails to be antisymmetric, so the result follows from the fact that, for any suitable \mathcal{G}' as above, we must have $\mathcal{S} \subseteq \mathcal{S}'$, and so if \mathcal{S}' is antisymmetric then so is \mathcal{S} . \square

We say a finite collection of weak schemas, $\mathcal{G}_1, \dots, \mathcal{G}_n$, is **compatible** if the relationship $(\mathcal{S}_1 \cup \dots \cup \mathcal{S}_n)^*$ is anti-symmetric (where $\mathcal{S}_1, \dots, \mathcal{S}_n$ are the specialization relations of $\mathcal{G}_1, \dots, \mathcal{G}_n$ respectively). Consequently we have, for any finite compatible collection of proper schemas, $\mathcal{G}_1, \dots, \mathcal{G}_n$, there exists a *weak schema merge* $\mathcal{G} = \bigsqcup_{i=1}^n \mathcal{G}_i$. Furthermore, since we define \mathcal{G} as the least upper bound of $\mathcal{G}_1, \dots, \mathcal{G}_n$, the operation is associative and commutative.

$\mathcal{G}_1 \sqcup \mathcal{G}_2$:

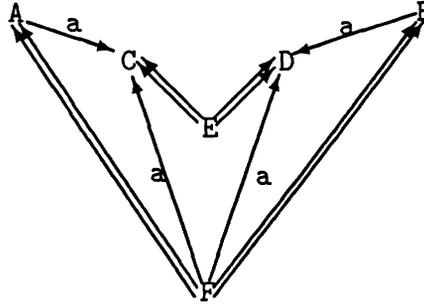


Figure 8: The least upper bound of two schemas

For example, the schemas \mathcal{G}_1 and \mathcal{G}_2 in figure 6 are compatible, and their weak schema merge is shown in figure 8.

4.2 Building proper schemas from weak schemas

We now must pay the price for our use of weak schemas: we must provide a way of introducing implicit classes into a weak schema \mathcal{G} , in order to form a proper schema $\overline{\mathcal{G}}$, such that if there are any proper schemas greater than \mathcal{G} then $\overline{\mathcal{G}}$ is such a schema.

First we introduce some new notation. For any $p \in C$ and any $a \in \mathcal{L}$, we write $\mathcal{R}(p, a)$ to denote the set of classes reachable from p via a -arrows

$$\mathcal{R}(p, a) = \{q \in C \mid p \xrightarrow{a} q\}$$

Further, for any set $X \subseteq C$, we use $\mathcal{R}(X, a)$ to denote the set of classes reachable from classes in X via a -arrows

$$\mathcal{R}(X, a) = \{q \in C \mid \exists p \in X \cdot p \xrightarrow{a} q\}$$

We define the function $Min_{\mathcal{S}} : \mathcal{P}(C) \rightarrow \mathcal{P}(C)$ so that, for any set $X \subseteq C$, $Min_{\mathcal{S}}(X)$ is the set of minimal elements of X under the ordering \mathcal{S} . That is

$$Min_{\mathcal{S}}(X) = \{p \in X \mid \forall q \in X \cdot \text{if } q \implies p \text{ then } q = p\}$$

where $\mathcal{P}(A)$ denotes the power set of the set A .

We now proceed to build a proper schema $\overline{\mathcal{G}} = (\overline{C}, \overline{\mathcal{E}}, \overline{\mathcal{S}})$ from \mathcal{G} as follows:

1. First we will construct a set, $Imp \subseteq \mathcal{P}(C)$, of sets of classes, corresponding to our implicit classes. We will construct Imp via a series of auxiliary definitions as follows:

$$\begin{aligned} I^0 &= \{\{p\} \mid p \in C\} \\ I^{n+1} &= \{\mathcal{R}(X, a) \mid X \in I^n, a \in \mathcal{L}\} \\ I^\infty &= \bigcup_{n=1}^{\infty} I^n \\ Imp &= \{Min_{\mathcal{S}}(X) \mid X \in I^\infty \text{ and } |Min_{\mathcal{S}}(X)| > 1\} \end{aligned}$$

Intuitively, Imp is the set of all sets of minimal classes which one can reach by following a series of arrows from some class in C , with cardinality greater than 1. Note that the process of forming Imp will halt since there are a finite number of subsets of $\mathcal{P}(C)$.

Using the example in figure 8,

$$\begin{aligned} I^0 &= \{\{A\}, \{B\}, \{C\}, \{D\}, \{E\}, \{F\}\} \\ I^1 &= \{\{C\}, \{D\}, \emptyset, \{C, D\}\} \\ I^2 &= \{\emptyset\} \\ I^\infty &= I^0 \cup I^1 \\ Imp &= \{Min_{\mathcal{S}}(\{C, D\})\} = \{\{C, D\}\} \end{aligned}$$

2. We define \overline{C} by first taking C and then adding a new class \overline{X} for every $X \in Imp$. That is

$$\overline{C} = C \cup \{\overline{X} \mid X \in Imp\}$$

Continuing with figure 8,

$$\overline{C} = \{A, B, C, D, E, F, \overline{\{C, D\}}\}$$

3. We define $\overline{\mathcal{E}}$ so that if $p \xrightarrow{a} q$ for each $q \in X$ then $p \xrightarrow{a} \overline{X} \in \overline{\mathcal{E}}$, while if there is a q such that $p \xrightarrow{a} q$ then $p \xrightarrow{a} q \in \overline{\mathcal{E}}$. Formally:

$$\begin{aligned} \overline{\mathcal{E}} = & \{x \xrightarrow{a} q \mid x \in \overline{C}, a \in \mathcal{L}, q \in \mathcal{R}(x, a)\} \\ & \cup \{x \xrightarrow{a} \overline{Y} \mid x \in \overline{C}, a \in \mathcal{L}, Y \in Imp, Y \subseteq \mathcal{R}(x, a)\} \end{aligned}$$

where $\mathcal{R}(\overline{X}, a) = \mathcal{R}(X, a)$ for all $X \in Imp$.

For our running example,

$$\overline{\mathcal{E}} = \mathcal{E} \cup \{F \xrightarrow{a} \overline{\{C, D\}}\}$$

4. We define $\overline{\mathcal{S}}$ by first taking \mathcal{S} and then adding every $\overline{X} \Longrightarrow \overline{Y}$ such that every class in Y has a specialization in X ; every $\overline{X} \Longrightarrow p$ where p has a specialization in X ; and every $p \Longrightarrow \overline{X}$ where p is a specialization of every class in X .

$$\begin{aligned} \overline{\mathcal{S}} = & \mathcal{S} \cup \{\overline{X} \Longrightarrow \overline{Y} \mid X, Y \in Imp, \forall p \in Y \cdot \exists q \in X \cdot q \Longrightarrow p \in \mathcal{S}\} \\ & \cup \{\overline{X} \Longrightarrow p \mid X \in Imp, p \in C, \exists q \in X \cdot q \Longrightarrow p \in \mathcal{S}\} \\ & \cup \{p \Longrightarrow \overline{X} \mid p \in C, X \in Imp, \forall q \in X \cdot p \Longrightarrow q \in \mathcal{S}\} \end{aligned}$$

Thus, for figure 8,

$$\overline{\mathcal{S}} = \mathcal{S} \cup \{ \overline{\{C, D\}} \Longrightarrow \overline{\{C, D\}} \} \cup \{ \overline{\{C, D\}} \Longrightarrow C, \overline{\{C, D\}} \Longrightarrow D \} \cup \{ E \Longrightarrow \overline{\{C, D\}} \}$$

Summarizing the effect on figure 8, we get the schema G3 in figure 7, with the class ? replaced by the class $\overline{\{C, D\}}$.

It can be shown that for any weak schema \mathcal{G} , $\overline{\mathcal{G}}$ is a weak schema and $\mathcal{G} \sqsubseteq \overline{\mathcal{G}}$. Furthermore, $\overline{\mathcal{G}}$ can be shown to respect condition 1 of the definition of a proper schema, and is therefore also a proper schema.

We would like to be able to show that $\overline{\mathcal{G}}$ is the least proper schema greater than \mathcal{G} . Unfortunately this isn't quite true. Firstly, it is possible to form other similar proper schemas by using different names for the implicit classes (compare this to alpha-conversion in the lambda calculus). Secondly, for any two sets $X, Y \in Imp$, if every class in Y has a specialization in X then our method will include the pair $(\overline{X}, \overline{Y})$ in $\overline{\mathcal{S}}$. However it is not necessarily the case that this specialization relation is required, and it might be safe to omit it. We

could attempt to modify our method so that such pairs are only introduced when required. Instead we will argue that, since the implicit classes have no additional information associated with them, it follows that these specialization relations do not introduce any extra information into the database schema, and so, since they seem natural, it is best to leave them there. Consequently we feel justified defining the **merge** of a compatible collection of database schemas, $\mathcal{G}_1, \dots, \mathcal{G}_n$, to be the database schema $\overline{\mathcal{G}}$, where $\mathcal{G} = \bigsqcup_{i=1}^n \mathcal{G}_i$.

Of course, not every merge of a collection of compatible schemas makes sense. That is, the new classes introduced may have no correspondence to anything in the real world. To capture this semantic aspect of our model, we would need to introduce a “consistency relationship” on \mathcal{C} , and require that, for every $X \in \text{Imp}$ and every $p, q \in X$, the pair (p, q) is in the consistency relationship. If this condition were violated, the schemas would be **inconsistent**, and $\overline{\mathcal{G}}$ would not exist. Note that checking consistency would be very efficient, since it just requires examining the consistency relationship. However, while the idea is interesting, it is beyond the scope of this paper. Suffice it to say that if the merge of $\mathcal{G}_1, \dots, \mathcal{G}_n$ fails, either because $\mathcal{G}_1, \dots, \mathcal{G}_n$ are *incompatible*, or because they are *inconsistent*, the merge should not proceed, and the user must re-assess the assumptions that were made to describe the schemas.

5 Cardinality Constraints and Keys

The model we have used so far concentrates on the semantic relationships between classes via specialization and arrow edges, but does not further describe arrows as participating in keys or having associated cardinality constraints. Cardinality constraints in the ER model are typically indicated on the edges between a relationship and an entity by labeling them “0..1”, “1..1”, “0..N”, or “1..N”, where “N” is used to indicate unrestricted upper bounds or “many” (rather than ∞ , which would be more appropriate).¹ For example, consider the **Lives** relationship between **Kennel** and **Dog** in Figure 1. As drawn, **Lives** is a “many-many” relationship, typically indicated by labeling the **home** and **occ** edges “0..N”; a dog may occupy many different kennels, and a kennel may be the home for many different dogs. If we decided to restrict **Lives** to indicate that a dog can live in at most one kennel the **occ** edge from **Lives** to **Dog** would be relabeled “0..1”. If we further decided that each dog *must* live in some kennel, we would change the **occ** label to “1..1”. In general, each edge a between a relationship R and entity E in an ER diagram means that in any instance of the ER diagram defining sets (extents) $\text{Ext}(E)$ and $\text{Ext}(R)$, $R \xrightarrow{a} E$ is *functional*. The edge

¹It is worth noting that there is little agreement on what edge labels to use, and what they mean in ER digrams, especially for ternary and higher degree relationships. No semantics are given in [9]. Introductory textbooks on databases avoid the question and merely give examples of binary relationships [10, 11, 12]; [13] is slightly more honest and says that “the semantics of ternary and higher-order relationship sets can become quite complex to comprehend.” Varying interpretations can be found in [14, 15, 3] We will follow the intuition given in [3] since it is one that we can understand.

label $\alpha.. \beta$ on a means

$$\forall e \in \text{Ext}(E). \alpha \leq |\{r | r \xrightarrow{a} e\}| \leq \beta.$$

While several extensions allow α and β to be any non-negative integers such that $\alpha \leq \beta$ (see [3]), we will restrict our attention to the more common cardinality constraints “0..1”, “1..1”, “0..N”, “1..N”.

As it stands, however, our model has no way of distinguishing these different edge semantics. Using the example of the previous paragraph, labeling the `occ` edge in the `Lives` relationship “0..1” rather than “0..N” could² result in the same graph in our model, *i.e.* the graph in Figure 2. In this section, we will capture such constraints by introducing “key constraints” on nodes, and argue that in some sense they are more general than the cardinality constraints typically found in ER models.

Key constraints, which indicate that certain attributes of an entity form a *key* for that entity, are another common form of assertions found in database models. As an example, in the ER and relational models, for the entity set `Person(SS#, Name, Address)`, we might claim that there are two keys: `{SS#}` and `{Name, Address}`. The intuition behind this statement is that if two people have the same social security number, or the same name and address, then they are the same person. Generalizing, one could claim that a set of *edges* of a relationship form a key for that relationship. As an example, for the `Lives` relationship in which all edges are marked “0..N”, we could claim that the `occ` and `home` edges form a key. In the terminology of proper schemas, we capture such key constraints by asserting that $\{a_1, a_2, \dots, a_n\}$ form a key for p , where each a_i is the label of some arrow out of p .

In general, a class may have several keys, or no key at all; the latter departure from a relational understanding of keys allows us to capture models in which there is a notion of object identity. A *superkey* of a class is any superset of a key. We may therefore think of the set of superkeys for a class p , $SK(p)$, as a set of sets of labels of arrows out of p . $SK(p)$ has the property that it is “upward closed”, *i.e.* if $s \in SK(p)$ and $s' \supseteq s$, then $s' \in SK(p)$.

We now have the constraint on specialization edges that if $p \implies q$ then $SK(p) \supseteq SK(q)$, *i.e.* all the keys for q are keys (or superkeys) for p . An example that satisfies this constraint can be found in Figure 9. In this example (borrowed from [3]), the assertion is made that the advisor of a student is a member of the thesis committee for that student: `Advisor` \implies `Committee`. Since the committee for a student consists of several faculty members, and each faculty member can be on several thesis committees, the set of keys for `Committee` is $\{\{\text{faculty}, \text{victim}\}\}$. However, since we assume that each student has at most one advisor, but that each faculty member can be the advisor of several students, the set of keys for `Advisor` is $\{\{\text{victim}\}\}$. Note that $\{\{\text{victim}\}, \{\text{faculty}, \text{victim}\}\} \supseteq \{\{\text{faculty}, \text{victim}\}\}$.

Our task now becomes to derive keys in the merged schema subject to this constraint.

²Of course, one might eliminate the `Lives` node entirely, and draw a single `home`-edge from `Dog` to `Kennel`, but this reasoning does not extend to ternary and higher degree relationships.



Figure 9: “Isa-A” Relation Between Relationships

Suppose schema \mathcal{G} is the proper schema merge of \mathcal{G}_1 and \mathcal{G}_2 . Each class p in \mathcal{G} appears at most once in each of \mathcal{G}_1 and \mathcal{G}_2 , with key assignments $\mathcal{SK}_1(p)$, $\mathcal{SK}_2(p)$ respectively (when defined). We define \mathcal{SK} to be a *satisfactory assignment* of keys to classes if

1. $\mathcal{SK}_1(p) \subseteq \mathcal{SK}(p)$, if $p \in \mathcal{C}_1$; and
2. $\mathcal{SK}_2(p) \subseteq \mathcal{SK}(p)$, if $p \in \mathcal{C}_2$; and
3. \mathcal{SK} satisfies the condition that $\mathcal{SK}(p) \supseteq \mathcal{SK}(q)$ whenever $p \implies q$.

It is readily checked that if \mathcal{SK} and \mathcal{SK}' are satisfactory assignments, then so is $\mathcal{SK} \cap \mathcal{SK}'$, defined by $(\mathcal{SK} \cap \mathcal{SK}')(p) = \mathcal{SK}(p) \cap \mathcal{SK}'(p)$. Thus there is a unique minimal satisfactory assignment to keys in classes.

Key constraints and cardinality constraints appear to be incomparable ideas. There are certain similarities: For example, the difference between labeling the edge `occ` in `Lives` “0..N” rather than “0..1” can be captured by stating that the set of keys for `Lives` with the “0..N” labeling is $\{\{\text{occ}, \text{home}\}\}$ as opposed to $\{\{\text{occ}\}\}$ for the “0..1” labeling. However, there are also dissimilarities. Keys do not seem to be able to capture all cardinality constraint labelings: For example, the labelings “0..1” and “1..1” (or “0..n”, “1..n”) on the `occ` edge cannot be distinguished by keys. On the other hand, cardinality constraints cannot capture all key assertions: For example, consider the relationship `Transaction` in Figure 10. The statement that `Transaction` has two keys, one being $\{\text{loc}, \text{amount}\}$, the other being $\{\text{card}, \text{amount}\}$, has no correspondence in terms of labeling edges.

Keys can also be used to determine when an object in the extent of a class in an instance one schema corresponds to an object in the extent of the same class in an instance of another schema. For example, if `Person` is a class in two schemas, \mathcal{G}_1 and \mathcal{G}_2 , which are being merged, and both schemas agree that $\{\text{SS}\#$ is a key for `Person`, then an object in the extent of `Person` in an instance of \mathcal{G}_1 corresponds to an object in the extent of `Person` in an instance of \mathcal{G}_2 if they have the same social security number. However, suppose that \mathcal{G}_1 claims that $\{\text{SS}\#$ is a key for `Person`, and \mathcal{G}_2 has an `SS#`-arrow for `Person` but does *not* claim that it is a key. Since $\{\text{SS}\#$ is a key for `Person` in the merged schema, an additional constraint has been placed on the extents of \mathcal{G}_2 : two objects in the extent of `Person` are

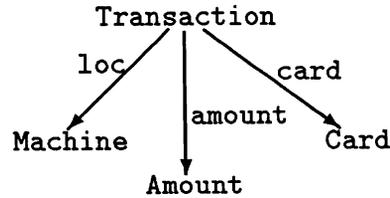


Figure 10: A Class with Multiple Keys

the same if they have the same social security number, no matter whether both are from an instance of \mathcal{G}_1 , both are from an instance of \mathcal{G}_2 , or one is from an instance of \mathcal{G}_1 while the other is from an instance of \mathcal{G}_2 . Furthermore, if \mathcal{G}_1 claims that $\{\text{SS}\#\}$ is a key for **Person** but \mathcal{G}_2 does *not have* an **SS#**-arrow for **Person**, then there is not way to tell when an object from the extent of **Person** in an instance of \mathcal{G}_1 corresponds to an object from the extent of **Person** in an instance of \mathcal{G}_2 .

6 Lower Merges

In Section 4 we defined the merge of a collection of schemas as their *least upper bound* under an information ordering: Whatever the individual schemas assert to be true is believed to be true in the merged schema. Thus, if we merge a collection of schemas, then any instance of the merged schema can be considered to be an instance of any of the schemas being merged. In some cases, however, it may be desirable to define the merge of a collection of schemas as their *greatest lower bound*: Whatever the individual schemas *agree on* as being true is believed to be true in the merged schemas. In this case any instances of the schemas being merged would also be instances of the merged schema, and, further, we would expect to be able to coalesce or take the union of a number of instances of the collection of schemas and use that as an instance of the merged schema. This kind of merge is likely to arise in, for example, the creation of a single schema from a collection of user views in database design.

We will refer to the merges defined in Section 4 as **upper merges**, and the formulation discussed in this section as **lower merges**.

As it stands, taking the lower bound of a collection of schemas using our information ordering is unsatisfactory since any information on which two schemas disagree on is lost. For example, if one schema has the class **Dog** with arrows **name** and **age**, and another has **Dog** with arrows **name** and **breed**, then in the lower bound of the two schemas the class **Dog** will only have the arrow **name**. What we want, however, is some way of saying that instances

of the class `Dog` *may* have `age`-arrows and *may* have `breed`-arrows, but are not necessarily required to do so. Worse still, if one schema has the class `Guide-Dog` and another does not, then the lower bound of the two schemas will not have `Guide-Dog`. The second problem can be dealt with easily by adding all classes involved in other schemas to each schema in a collection before proceeding with the construction of the lower merge. The first problem, however, is more difficult and requires us to extend our definition of (weak) schemas.

We define the semi-lattice of **participation constraints**, ordered by \leq , to be as shown in figure 11. We will extend the definition of (weak) schemas by associating a participation

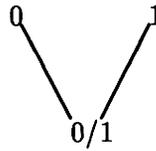


Figure 11: The semi-lattice of participation constraints

constraint with each arrow of a schema. The idea is that, if a class p has an a -arrow of class q , then if the arrow has participation constraint 1 then every instance of class p *must* have a an a -arrow to an instance of class q ; if the arrow has participation constraint 0/1 then an instance of p *may* have an a -arrow of class q ; and if the arrow has constraint 0 then an instance of p *does not* have an a -arrow of class q . We adopt the convention of not drawing arrows with the participation constraint 0 in our diagram, and, further, assume that a schema which does not have some arrow $p \xrightarrow{a} q$ is equivalent to the same schema but with the arrow $p \xrightarrow{a} q$ with participation constraint 0.

Now, if one schema has an arrow which is not included in another, then we can assume that the second schema also has the arrow with participation constraint 0, and we can take the greatest lower bound of the participation constraints (under the ordering \leq) to be the participation constraint of the arrow in the merged schema.

Hence, with the addition of participation constraints, we can form the *weak lower merge* of a collection of schemas in a similar manner to that used to construct the *weak upper merges* in Section 4. We can also build a proper schema from a weak lower merge using an algorithm similar to that in Section 4, except that that the implicit classes are introduced above, rather than below, the sets of proper schemas that they represent.

It is worth noting that, while upper and lower merges represent two extreme views of what the merge of a collection of schemas should be, there may well be valid and useful concepts of merges lying in between the two. However the authors believe that, in order for a concept of a merge to be valid and well defined, it should have a definition in terms of an information ordering similar to the ones given here.

7 Conclusions

Using a simple but general formalism, we have characterized the *weak schema merge* of a collection of schemas as their least upper bound. The *merge* of these schemas is then defined by translating the weak schema merge to a *proper* schema. The translation introduces new “implicit” classes as required, and identifies their origin in their name. Although not discussed in detail in this paper, the “real-world” validity of an implicit class can be efficiently checked by consulting a consistency relationship between the classes from which the implicit class was formed.

Despite the simplicity of our mathematical construction, we believe that using an information ordering is the right way of describing the merge of schemas: it has a well-defined result, and the merge operation is associative and commutative. Thus user assertions about the relationships between schemas can be thought of as real assertions rather than “guiding heuristics” since the merge is independent of the order in which the assertions are stated. The approach in this paper focused on the upper merge of schemas, which seems to be a natural interpretation for merging pre-existing, heterogeneous databases. Other interpretations of the merge can be used, including the lower merge, by varying the information ordering used.

The approach presented in this paper can be generalized to describe the merge in a number of other data models by representing schemas in other data models as “restricted” instances of schemas in our general model (*i.e.* stratifying classes in terms of their meaning in other models), and finding their proper schema merge. Our merge can be shown to “preserve strata”, guaranteeing that the result will be an instance of the original model; a proof of this with full details can be found in [5].

To use this approach as a practical schema merging tool, several issues should be addressed. Firstly, more attention should be paid to how cardinality constraints should be incorporated. While our preliminary approach has been to use a notion of keys, other ideas include allowing arrows to be “multivalued functions” as in [2]; [5] shows how this idea can be extended to our model. We also have not adequately captured the lower bound of “0” on arrow edges (also known as “total versus partial participation” in relationships). Secondly, some form of assistance should be given for “restructuring” schemas to obtain a better merge. Not only can “naming” conflicts occur (such as homonyms and synonyms), but “structural” conflicts can occur. For example, an attribute in one schema may look like an entity in another schema, or a many-one relationship may be a single arrow in one schema but introduce a relationship node in another schema. In these cases, the merge will not “resolve” the differences but present both interpretations. To force an integration, we need some kind of “normal form”. Thirdly, we need to evaluate how many implicit classes can be introduced in the merge. Although in the examples we have looked at this number has been small, it may be possible to construct pathological examples in which the number of implicit classes is very large; however, we do not think these are likely to occur in practice.

Fourthly, we must discuss how to merge instances; for a discussion of the problems involved, see [16].

We have found that the simplicity of the method and presence of strong theoretical underpinnings have made extensions of the technique very easy to develop. In addition, we have been able to rapidly prototype the method, together with a graphical interface for creating and displaying schema graphs.

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