

**Semantics for Update Rule Programs and Implementation  
in a Relational Database Management System <sup>1</sup>**

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**Abstract**

In this paper we present our research on defining a correct semantics for a class of update rule (UR) programs, and discuss implementing these programs in a DBMS environment. Update rules execute by updating relations in a database which may cause the further execution of rules. A correct semantics must guarantee that the execution of the rules will terminate and that it will produce a minimal updated database. The class of UR programs is syntactically identified, based upon a concept that is similar to stratification. We extend the strict definition of stratification and allow a relaxed criterion for partitioning of the rules in the UR program. This relaxation allows a limited degree of non-determinism in rule execution. We define an execution semantics based upon a monotonic fixpoint operator  $T_{UR}$ , resulting in a set of fixpoints for UR. The monotonicity of the operator is maintained by explicitly representing the effect of asserting and retracting tuples in the database. A declarative semantics for the update rule program is obtained by associating a normal logic program  $\overline{UR}$  to represent the UR program. We use the stable model semantics which characterize a normal logic program by a set of minimal models which are called stable models. We show the equivalence between the set of fixpoints for UR and the set of stable models for  $\overline{UR}$ . We briefly discuss implementing the fixpoint semantics of the UR program in a DBMS environment. Relations that can be updated by the rules are *updatable* relations and they are extended with two flags. An update rule is represented by a database query, which queries the updatable relations as well as database relations, i.e., those relations which are not updated by rules. We describe an algorithm to process the queries and compute a fixpoint in the DBMS environment and obtain a final database.

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## 1. Introduction

Since the introduction of deductive databases as a research area [Bocc86a, Bocc86b, KoGM87, Mink88, MUvG86, TsZa86], there has been a corresponding interest in extending the functionality of database management systems (DBMS) to provide support for rules. Some of this research has been described in [CeWi90, DeEt88, MaSi88, SeLR88, SeLR93, SiMa88, Wido91, WiFi90]. Whereas rules in deductive database systems have been primarily perceived as supporting retrieval queries against the database, most DBMS research has focused on supporting rules whose execution can update the database and trigger the further execution of rules. This paradigm is similar to the forward-chaining, rule-based production system paradigm in artificial intelligence research [Haye75]. The motivation for choosing this paradigm is partly because triggers and integrity constraints, which are common in DBMS, also update the database, and are often implemented in a similar fashion to forward-chaining rules.

The success of research in deductive bases, where Horn logic programs are integrated with function-free first-order relational databases, can be largely attributed to the fact that there is a strong mathematical foundation for deductive databases. Horn logic programs have a declarative semantics and an equivalent fixpoint semantics [vEko76]. However, the same is not true for database updates and rules whose execution may update the database. There has been some research in the area of precisely defining the semantics associated with updates. In [FKUV86], a framework for dealing with the general problem of database updates is presented. Since then, there has been research in updates in logic programs [MaWa88, NaKr88] where the semantics is defined using a Dynamic Logic, and there has been research in very powerful procedural languages that are equivalent to Datalog-like extensions with fixpoint semantics [AbSV90, AbVi90, AbVi91, SiMa88]. Most DBMS implementations of production rules [DeEt88, MaSi88, SeLR93, SiMa88, WiFi90] have an operational or execution semantics defined for them.

In this paper, we are interested in defining a semantics for update rules in a DBMS, where rule execution can update the database and cause further execution of rules. Our research focuses on supporting update rule programs (a collection of update rules and an extensional database of facts) in an extended DBMS environment. Although there has been previous research in the general area of providing a semantics for updates in a logic program or in a DBMS, and in defining powerful rule languages for updates, as mentioned, this problem has not been studied in the context of DBMS implementations. Our motivation is to define a class of update rule programs which can be identified in a fairly straightforward manner by a DBMS designer, and which would be useful in DBMS applications such as to capture triggering information, to maintain integrity constraints, or to make simple inferences from the information in the database. An important criterion in defining this class of update rule programs is that its semantics must be easily, and correctly, implemented in a relational DBMS with some extensions.

There are two problems that must be resolved in providing a correct execution semantics for update rules in a DBMS. One problem is to deal with non-terminating execution of update rule programs where rules insert and/or delete the same tuples indefinitely. The second problem arises when there are negative literals in the rules. With different execution schedules that execute rules in different sequences, a database produced by some schedule may not be minimal. Informally, this means a database produced by one schedule may be a subset of a database produced by another schedule. This occurs because negative literals that are evaluated by the rules may be affected by subsequent insertions and/or deletions of tuples, by other rules.

To deal with the first problem of non-terminating programs, we define a fixpoint semantics for update rule programs based on a monotonic fixpoint operator. The monotonicity of the fixpoint operator is achieved by explicitly recording the effects of inserts and deletes of tuples in the database. The retraction of a tuple (by a rule) is not affected by subsequent assertions of the same tuple (by other rules), and this ensures the termination of the execution. The second problem of how to deal with negative literals has been researched within the context of logic programs. Stratification [ApBW88] is a technique used to identify a special class of logic programs where the rules have negative literals. Stratification imposes an ordering between the rules and such an ordering supports a correct interpretation of the negative literals in the rules. In previous research [Rasc90, Rasc94], we defined a strict

stratification criterion for the update rules. However, stratification is an overly strict criterion, and there are many update rule programs which, while not stratified, capture some very useful properties in a DBMS [RaLo94]. A further restriction of stratification is that the execution of the program must be deterministic, i.e., it must produce a single answer. However, there are many programs that are naturally non-deterministic. For example, when the database does not satisfy a given integrity constraint, then there may be several update rules which could be applied to restore consistency. These rules could be non-deterministically chosen, depending on the application, and could result in different answers.

In this paper, we relax the strict definition of stratification for update rules, which was first introduced in [Rasc90, Rasc94]. We present a more flexible definition for partitioning the update rule program  $UR$  into a set of partitions,  $UR_1 \cup \dots \cup UR_n$ . Each partition may include cliques of rules which do not satisfy the criterion for strict stratification. As with strictly stratified programs, there is an implicit ordering among the update rules from one partition to the next. We define a fixpoint semantics for the  $UR$  program based upon a fixpoint operator  $T_{UR}$ . Rules in partition  $UR_i$  must be executed and a fixpoint obtained before executing rules in the partition  $UR_{i+1}$ . Since rules are non-deterministically selected for execution, there are several possible fixpoints. Each fixpoint for the  $UR$  program corresponds to a final database. The monotonicity of the fixpoint operator, the condition that once a tuple is retracted it is not affected by subsequent assertions (of the same tuple) and the stratification criterion results in an execution which is guaranteed to terminate and the execution will also produce a minimal database.

To provide a precise definition for the final (minimal) database(s), we associate a normal logic program  $\overline{UR}$  with each  $UR$  program. The translation to obtain this program  $\overline{UR}$ , from  $UR$ , interprets rules which update the database by retracting tuples as integrity constraints. This is different from other approaches, where retractions in the heads of rules are treated as negative literals in the heads of the rules. We believe our interpretation is more natural for DBMS applications, and we discuss this in the section on related research. We use the stable model semantics for normal logic programs [GeLi88] which characterizes a normal logic program with a set of minimal models called *stable models*. A stable model for the program  $\overline{UR}$  must be identical to a final database obtained when a fixpoint is reached from executing the rules of  $UR$ , and vice versa. The translation to obtain  $\overline{UR}$  and the declarative semantics for  $UR$ , based on stable models is described in detail in this paper.

Implementing the  $UR$  program in a DBMS is straightforward and computing the fixpoint for the rules is similar to query processing in a DBMS. Each predicate in the  $UR$  program corresponds to a relation. The *updatable* relations that are updated by the rules are extended with two additional attributes. This extension allows us to explicitly store information on tuples which are inserted or deleted by the rules. The rules themselves are translated into queries which retrieve tuples from the updatable relations and the database relations (which are not updated by the rules). Queries corresponding to the rules in each partition are executed, in turn, until the queries in the partition can no longer update the updatable relations. We note that the complexity of query evaluation is not affected due to the additional attributes used with the updatable relations. The simplicity of the implementation motivates the choice of this criterion for relaxing the strict stratification when defining the acceptable update rule programs.

This paper is organized as follows: In section 2, we identify the syntax of the  $UR$  programs and provide some example programs to illustrate the problems that arise during execution of  $UR$  programs. In section 3, we introduce the concept of stratification and define  $UR$  programs which are strictly stratified. We then describe two simple extensions to the stratification conditions. Each extension introduces *cliques* of rules (which fail the strict stratification criterion) into each partition, and provides a more flexible criterion for partitioning the rules of the  $UR$  program. The usefulness of these extensions is motivated with example  $UR$  programs. In section 4, we define the fixpoint semantics for the  $UR$  programs defined in section 3. In section 5, we provide a translation to obtain a normal logic program  $\overline{UR}$ , and we present the equivalent declarative semantics based on the stable model semantics for  $\overline{UR}$ . In section 6, we provide a proof that the fixpoint semantics for  $UR$  is sound and complete wrt the declarative

semantics for  $\overline{\text{UR}}$ . In section 7, we discuss related research on providing a semantics for updates, and we compare these approaches with our research on semantics of UR programs. In section 8, we outline a method for implementing the fixpoint semantics of the UR programs (described in section 4) in a relational DBMS. Finally, in section 9, we summarize our research and we discuss future research. Appendix 1 provides a translation from an update rule program to a normal logic program in the first order case.

## 2. A Syntax for Update Rule Programs

In this section we introduce the syntax for specifying update rules. We borrow concepts from the OPS5 production system language [Forg81 and Forg82], as well as the Datalog language with procedural extensions [AbVi90, AbVi91].

### 2.1. Syntax for Update Rules

An *update rule* consists of (1) the **name** (2) the **antecedent** on the left hand side (LHS), also referred to as the body of the rule, (3) the symbol  $\rightarrow$ , and (4) the **consequent actions**, on the right hand side (RHS), also known as the head of the rule.

The antecedent is a conjunction of first order positive literals of the form  $P(\bar{u})$  or negative literals of the form  $\neg Q(\bar{v})$  or *evaluable* predicates [Ullm89] of the form:  $\text{eval}(\text{op}, \text{arg}_1, \text{arg}_2)$ . The **consequent actions** are of the form **assert**  $R(\bar{u})$  or **retract**  $S(\bar{v})$ , where the **assert** and **retract** are special constants reserved for the assert and retract operations (to be explained later) and  $R$  and  $S$  are predicates. Let  $P, Q, R, S$ , etc., be predicates which are represented by relations in the DBMS. The relations corresponding to predicates that occur in the actions in the heads of the update rules are *updatable* relations, e.g.,  $R$  or  $S$ . Relations that correspond to predicates that only occur in the body of update rules, but not in the heads of rules are database relations. They are not updated by the update rules but they can be updated by explicit insertions and deletions, just as the updatable relations. Literals in the body of a rule can refer to either database or updatable relations.  $\bar{u}$  and  $\bar{v}$  are vectors of terms from a non-empty finite or infinite set of constants  $C$ , and from a set of variables  $X$ . An *evaluable predicate* *eval* is a binary built-in predicate supported by many database systems [Ullm85]; *op* is one of the binary comparison operators such as  $\{<, \leq, >, \geq, =, \neq\}$  and  $\text{arg}_1$  and  $\text{arg}_2$  are either variables or constants from the set  $C$ . We assume that there exists a total order in the elements of  $C$ <sup>1</sup>.

We assume that all variables are *range-restricted* [Demo82, Nico82, NiDe83], i.e., any variable that occurs in a literal, or in the evaluable predicate *eval*, must appear in a positive literal in the body of a rule. This restriction correspond to the safety of evaluating queries<sup>1</sup>. It ensures that only ground atoms are inserted into the updatable relations through the execution of the actions of the update rules. The syntax restricts the literal referred to by the **retract** action, to occur positively in the antecedent of that rule, so that only ground atoms (tuples) that are actually in the database, (in the updatable relations) are retracted through the execution of the update rules.

A *function-free update rule program*  $\text{UR}$ , consists of the following<sup>2</sup>:

<sup>1</sup> Note that the set of constants usually corresponds to a collection of sets of different types, e.g., integer, string, etc. To simplify the discussion, we assume all the constants to be of the same type.

<sup>1</sup> Note that the variables are not also instantiated to safely computable values corresponding to either arithmetic functions or aggregate functions, as is common in a DBMS. It is straightforward to modify the syntax for the rules to include such safely computable functions, corresponding to the variables occurring in a predicate.

<sup>2</sup> Note that an action in the head of a rule which updates the value(s) of one (or more) variable(s) of a literal (i.e., attributes of a corresponding tuple), can be interpreted as a pair of corresponding **assert** and **retract** actions. Such an update rule can be rewritten as a pair of appropriate a-update and r-update rules, relevant to this literal.

- (i) A set of update rules, each of which has a *single*<sup>3</sup> **assert** action in its head. These are *a-update* rules and they *assert* the literal occurring in the head of the rule, or they insert tuples into the corresponding updatable relation.
- (ii) A set of update rules, each of which has a *single*<sup>3</sup> **retract** action in its consequent. These are *r-update* rules and they *retract* the literal which occurs in the head of the rule, or they retract tuples from the corresponding updatable relation.
- (iii) An initial extensional database of ground atoms  $EDB_{init}$ .

We informally describe the execution of an update rule program. The antecedent of each rule is interpreted as a query against the relations, as follows: (1) For each of the positive literals,  $P(\bar{u})$ , relation  $P$  is queried, and a set of *instantiated* tuples of  $P$  satisfying each positive literal in the antecedent is retrieved. (2) For each of the negative literals,  $\neg Q(\bar{v})$ , the query is verified against the corresponding relation  $Q$ . Each variable  $x$  in  $\bar{v}$  is range restricted to the value obtained by evaluating a query for some positive literal  $P$  in which  $x$  occurs. (3) Finally, for each occurrence of the evaluable predicate, *eval*, it is evaluated by the DBMS to return a value which is either true or false. Since the variables in the *eval* predicate are range-restricted they can be evaluated safely; this is discussed later. The antecedent of an update rule is *satisfied* if the relations contain instantiated tuples corresponding to each of the positive literals, if the relations do not contain tuples corresponding to the negative literals, and if all occurrences of the *eval* predicate in the body of the rule evaluate to true. Rules whose antecedents are satisfied are (non-deterministically) selected for execution and they update the database. Depending on the action in the head of the selected rule, the updatable relations are updated. Either new facts are asserted by the *a-update* rules or existing facts are retracted by the *r-update* rules. This process continues until the updatable relations can no longer be updated.

## 2.2. Problems in Executing Example Update Rule Programs

There are several problems that may arise when executing a set of update rules in a DBMS. If we interpret the **assert** and **retract** actions in a straightforward manner by inserting and deleting database tuples from the corresponding relations, then there is a possibility that the execution of the program may not terminate and the relations may be updated indefinitely. The second problem is that when there is more than one execution schedule, the execution may not produce a minimal database. We present some motivating example programs that highlight these problems.

### Example 1

Consider the update rule program with relations *Employee*, *GoodWorker*, *Manager* and *Unfriendly*, where the initial database has the tuples, {Employee(Mike), GoodWorker(Mike).}, and with the following set of update rules:

$r_1$ : Employee(X), GoodWorker(X)  $\rightarrow$  **assert** Manager(X)

$r_2$ : Employee(X),  $\neg$  HasOffice(X)  $\rightarrow$  **assert** Unfriendly(X)

$r_3$ : Manager(X), Unfriendly(X)  $\rightarrow$  **retract** Manager(X)

The first rule promotes employees who are good workers to managers, while the second rule asserts the fact that employees who do not have offices become unfriendly. Finally the third rule rescinds the promotion of managers who are unfriendly. Given this initial database and set of rules, the rules will execute until a fixpoint is reached and the database can no longer be updated. Rules  $r_1$ , and  $r_2$  will execute (in any order) and the tuples

<sup>3</sup> An update rule program in which a rule can have multiple actions in the head can be replaced with an equivalent update rule program where each rule has a single action in the head. This may require the introduction of special predicates, in the new program, which are previously unused in the original program.

Manager(Mike) and Unfriendly(Mike) will be added to the database. Next, the rule  $r_3$  executes and the tuple Manager(Mike) will be deleted from the the database. Subsequently,  $r_1$  and  $r_3$  will execute, first inserting the tuple Manager(Mike) and then deleting this tuple. Processing of these two rules could continue indefinitely, alternately inserting and deleting the tuple Manager(Mike) from the Manager relation, and the UR program may not be able to terminate.

### Example 2

This is an example where the update rules produce different databases, due to differences in selecting and executing the rules.

The initial database = {Employee(Mike), GoodWorker(Mike)} and the update rules are as follows:

$r_1$ : Employee(X), GoodWorker(X)  $\rightarrow$  **assert** Manager(X)

$r_2$ : Employee(X), Manager(X)  $\rightarrow$  **assert** IncreasePay(X)

$r_3$ : Employee(X),  $\neg$  Manager(X)  $\rightarrow$  **assert** DecreasePay(X)

The first rule promotes employees who are good workers to managers. The second rule increases the salary of managers, while the third rule decreases the salary of employees who are not managers. The problem here is that depending on the order in which the rules are selected for execution, there will be two different answers obtained. For example, if the update rules are executed in the sequence  $r_1$  followed by  $r_2$ , then Mike will be promoted to a manager and his salary will be increased. However, if the execution sequence was  $r_3$  followed by  $r_1$  and  $r_2$ , then, first, Mike's salary will be decreased, following which Mike will be promoted to a manager, and his salary will be increased. It is in this sense that we say that the second execution is not minimal or does not produce a minimal database compared to the first execution. Informally, the database resulting from the asserts and retracts of the executed rules, produced by one execution schedule, includes all the answers corresponding to a different execution. The formal definition of a minimal database will be introduced in section 6. We will define the equivalent declarative stable model semantics, and the corresponding (minimal) stable models which correspond to the minimal databases.

Example 2 also points out the difference between a deterministic and a non-deterministic execution. Suppose we were to consider a deterministic rule execution strategy that allowed all the rules to execute simultaneously. Then, the rules  $r_1$  and  $r_3$  would execute and Mike would be promoted to a manager and his salary would be decreased simultaneously. Then rule  $r_2$  would execute and his salary would be increased. Such an execution is deterministic in that there can be only one final answer.

Another variation results when we consider the execution of a single rule. Consider the following rule:

$r$ : ManagerOf(P, X), ManagerOf(P, Y), Employee(P)  $\rightarrow$  **retract** ManagerOf(P,X)

Suppose Mary is an employee with two managers Jane and Martha. Then, the set of tuples {ManagerOf(Mary, Jane), ManagerOf(Mary, Martha)} will satisfy the literals ManagerOf(P,X) and ManagerOf(P,Y). A deterministic execution will then delete both the tuples of ManagerOf leaving Mary without a manager! However, a non-deterministic execution will only delete one of those tuples. These issues relating to deterministic or non-deterministic execution and tuple-oriented or set-oriented execution have been studied in different contexts. Researchers who study theoretical issues have reported on such executions in [AbVi90, AbVi91] and it has also been studied from an implementation viewpoint since the set-oriented execution can provide more efficient execution [PaGo91, WiFi91]. We will discuss this issue in a later section where we compare our research with other research.

### 3. Identifying Acceptable Update Rule Programs

In the previous section, we saw examples of update rule programs whose execution did not terminate, or did not produce minimal databases. To avoid this problem, we require a method to syntactically identify legal update



rule programs and define a semantics for these programs that disallows this behavior. One reason for the undesirable behavior of programs is that negative literals in the body of update rules must be correctly interpreted, as the tuples are asserted or retracted. Informally, an error occurs if a literal occurring negatively in a rule is interpreted to be false, while there is another rule which can assert a tuple which can prove this literal. Similarly, an error may occur if a literal occurring negatively in the body of a rule is interpreted to be true, while there may be another rule which can retract some tuple so that after the retraction, the literal is no longer true. In order to solve this problem, we apply the concept of stratification to partition the rules of a update rule program. The order imposed by this partitioning of the rules will order the sequence of execution of the update rules. As a result, the literals will be interpreted correctly, as defined by the corresponding fixpoint semantics for the update rule program.

Stratified logic programs are an extension of Horn programs to more general logic programs to allow negative literals in the antecedent of a rule. Since we apply the concept of stratification in our research, we informally define it in this section, as it applies to a stratified logic program [ApBW88]. We refer the reader to [ApBW88] for a discussion on stratified logic programs.

A logic program  $P$  consists of a finite set of rules of the following form:

$$A \leftarrow L_1, L_2, \dots, L_m$$

where  $A$  is an atom and each of the  $L_i$  are literals. If  $m=0$ , then  $A$  is a fact.

A logic program  $P$  is stratified if there exists a *partition*  $P = P_1 \dot{\cup} \dots \dot{\cup} P_n$  where  $\dot{\cup}$  is a disjunctive union, such that the following conditions hold for  $i = 1, 2, \dots, n$ :

- (1) for each predicate symbol, corresponding to a literal occurring *positively* in the body of a rule in  $P_i$ , its definition is contained within  $\bigcup_{j \leq i} P_j$ . The definition (of a predicate symbol) is all rules in which the predicate symbol corresponds to the literal occurring positively in the head of the rule.
- (2) for each predicate symbol, corresponding to a literal occurring *negatively* in the body of a rule in  $P_i$ , then its definition is contained within  $\bigcup_{j < i} P_j$ .

$P_1$  may not contain any rules of  $P$ . We say that  $P$  is stratified by  $P_1 \dot{\cup} \dots \dot{\cup} P_n$  and each  $P_i$  is called a stratum of  $P$ .

### 3.1. A Stratified UR program

#### Definitions

An update rule is an *a-update* rule if it has a single **assert** action in its head.

An update rule is an *r-update* rule if it has a single **retract** action in its head.

A rule is *relevant* to a predicate  $Q$  if the literal  $Q$  occurs in the head of the rule.

A rule  $p$  has a *positive/negative dependency* on the predicate  $Q$  when  $Q$  occurs in a positive or negative literal in the body of the rule. The predicate  $Q$  is a distinguished predicate in the rule.

A predicate  $P$  is retractable if there is an *r-update* rule relevant to that predicate, otherwise the predicate is unretractable. □

A stratified update rule program  $UR$ , is a function-free stratified program and comprises the following:

- (1) a set of *a-update* rules, as follows:

$r_1: A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow \mathbf{assert} P$

(2) a set of r-update rules, as follows:

$r_2: P, C_1, \dots, C_c, \neg D_1, \dots, \neg D_d \rightarrow \mathbf{retract} P$

(3) an initial database of facts ( $EDB_{init}$ ).

Note that we have not considered the evaluable predicates in the body of the rules. This is because the presence of the evaluable predicates does not affect the stratification criterion for the program.

There must exist a partition so that UR is a stratified program. Thus,  $UR = UR_0 \dot{\cup} UR_2 \dots \dot{\cup} UR_n$ . Each of the partitions  $UR_i$  comprises a set of a-update rules and a set of r-update rules; the sets of a-update rules or r-update rules in each partition may be empty. Partition  $UR_0$  corresponds to the initial database  $EDB_{init}$  and there are no rules in  $UR_0$ . The following conditions must hold for the stratification of the UR program:

- (1)  $UR = UR_0 \dot{\cup} UR_1 \dot{\cup} \dots \dot{\cup} UR_n$
- (2) For every distinguished predicate, occurring in a positive literal  $A_k$  or  $C_k$ , in a rule such as  $r_1$  or  $r_2$  in  $UR_i$ , all relevant a-update rules (in which the predicate occurs in the **assert** action, in the head), must be contained within  $\bigcup_{j \leq i} UR_j$ .
- (3) For every distinguished predicate, occurring in a positive literal  $A_k$  or  $C_k$ , in a rule in  $UR_i$ , all relevant r-update rules (where the predicate occurs in the **retract** action) must be contained within  $\bigcup_{j < i} UR_j$ .
- (4) For every distinguished predicate, occurring in a negative literal  $C_k$  or  $D_k$ , in a rule in  $UR_i$ , all relevant a-update rules must be contained within  $\bigcup_{j < i} UR_j$ .
- (5) For every distinguished predicate, occurring in a negative literal  $C_k$  or  $D_k$ , in a rule in  $UR_i$ , all relevant r-update rules must be contained within  $\bigcup_{j < i} UR_j$ .
- (6) Finally, for the distinguished predicate occurring in a positive literal  $P$ , in an r-update rule such as  $r_2$ , which has **retract**  $P$  in its head, in  $UR_i$ , all relevant rules with  $P$  in the head must be contained within  $\bigcup_{j \leq i} UR_j$ .  $\square$

Note that condition 6, together with the syntactic restriction that the distinguished predicate  $P$  must occur in a positive literal in an r-update rule such as  $r_2$  that has (**retract**  $P$ ) in its head, will place *all* r-update rules with (**retract**  $P$ ) in their heads in the same partition. In addition, it will not admit r-update rules with **retract**  $P$  in the head, and where some other distinguished predicate  $A_k$  or  $B_k$  is also  $P$ .

Based on this stratification condition, we note that in Example 2, the update rule  $r_3$  will be placed in a higher partition w.r.t. the other two rules. The impact of this partitioning on the execution of the rules is to ensure that a minimal database is obtained. The proof of obtaining a minimal database for the strictly-stratified programs is in a paper on related research [Rasc94].

### 3.2. Relaxing the Conditions for Stratification

Enforcing that a UR program is stratified is a very strict condition which eliminates many useful programs. When integrity constraints are maintained by r-update rules, or when a-update rules deduce new information, these rules query the database and make conditional updates. The queries and conditional updates made by two or more



rules may violate the strict stratification criterion previously described. It would therefore be advantageous to relax the stratification criteria so as to accept more UR programs. However, the relaxed criterion must still ensure that the execution of programs terminates and produces minimal databases. If we consider the problem of combining multiple knowledge bases, and answering queries and maintaining a combined knowledge base, then it becomes very important that the strict stratification criterion be relaxed. We motivate the relaxations by presenting example UR programs not meeting the strict stratification criteria. Nonetheless, they are useful programs. We also note that the strictly-stratified programs are deterministic and compute a single answer; this can be a drawback in many instances.

### 3.2.1. First Relaxation

We first relax the stratification conditions to allow r-update rules that are relevant to some set of predicates to co-exist in the same partition when they all have a mutual positive dependency on all the predicates in that set. This exception allows us to include rules which are illegal under the initial strict definition of stratification. Consider a database with relations corresponding to the relations *person*, *male* and *female*. A person can be either a male or a female, but not both. This is a constraint that we want to be enforced. The following pair of r-update rules maintain this constraint.

#### Example 3

$r_1$ :  $\text{person}(x), \text{male}(x), \text{female}(x), \text{choice}(x, "m") \rightarrow \text{retract } \text{female}(x)$

$r_2$ :  $\text{person}(x), \text{male}(x), \text{female}(x), \text{choice}(x, "f") \rightarrow \text{retract } \text{male}(x)$

In the event that the database is updated so that it violates this constraint, then, either the fact that a person is a male or that she is a female must be retracted from the database, i.e., the corresponding tuple must be retracted. Since this is not an arbitrary decision which could update the database in a non-deterministic manner, in this example, we introduce an additional predicate *choice*. This predicate is used to prompt the user to select whether the person is a male or a female. Depending upon the value of (the second attribute of) *choice*, either the tuple satisfying *female*(*x*) or *male*(*x*), for some bound value of variable *x*, will be retracted from the corresponding relation.

If we apply the strict definition of stratification to these rules, condition (3) of this definition in section 3.1 requires that rule  $r_1$  be in a lower partition than rule  $r_2$  and that  $r_2$  be in a lower partition than  $r_1$ . Obviously, these two requirements cannot be simultaneously satisfied, and the program would be rejected.

This situation is characterized by two or more rules querying the same subset of predicates in the body of the rule(s), or querying some mutually overlapping subsets of some set of predicates. The update actions of the rules are different, and the rules are designed so that usually all the rules are not expected to be executed. Typically, this situation occurs when an integrity constraint is being verified against the database. When the database violates the constraint, there may be many different ways to restore consistency, corresponding to each rule. One of the ways is chosen, either selected by the user as in this example, or more non-deterministically, when a particular rule is selected by the system. Thus, these rules will check the same conditions in the body of the rules, while the actions in the head of each rule will be different. As a result, the rules will fail the strict-stratification criterion.

#### Example 4

Consider another example where the rules do not satisfy the strict stratification criterion but are still useful. In this example, two different answers can be non-deterministically obtained by executing the rules. The database has relations corresponding to the predicates *manager*, *goodworker* and *unfriendly*, and the rules are as follows:

$r_1$ :  $\text{manager}(x), \text{unfriendly}(x) \rightarrow \text{retract } \text{manager}(x)$

$r_2$ :  $\text{manager}(x), \text{unfriendly}(x), \text{goodworker}(x) \rightarrow \text{retract unfriendly}(x)$

$r_3$ :  $\text{unfriendly}(x), \text{goodworker}(x) \rightarrow \text{retract goodworker}(x)$

Consider Joan who is a good worker and a manager, but is unfriendly. We can either execute rules  $r_1$  and  $r_3$  which maintain the constraints that an unfriendly person should not be a manager, and that an unfriendly person is not a good worker, respectively. The final database is Joan is an unfriendly person. Alternately, we can execute rule  $r_2$  which maintains the constraint that a person who is a manager and also a good worker is not an unfriendly person. The final database is that Joan is a good worker and a manager.

In these cases, instead of rejecting the program, our solution is to relax the strict stratification criteria, so that these programs are accepted. The relaxation to accept rules such as those in Examples 3 and 4 applies to condition (3) of the stratification criterion. We first identify a *clique* of r-update rules that are relevant to a set of predicates, say  $\{P_1, \dots, P_n\}$ . These r-update rules in the clique may co-exist in the same partition when they each have mutual positive dependencies on these same distinguished predicates,  $P_1, \dots, P_n$ .

### Definition

A set of r-update rules participate in an *r-clique* if there is some *maximal* set of predicates  $\{P_1, \dots, P_n\}$  such that each rule in the r-clique is relevant to some predicate from this set; if there is at least one rule in the r-clique which is relevant to each predicate in this set; and if all of the predicates in the set are distinguished and occur in a positive literal in each rule of this r-clique. All predicates in this maximal set are *retractable* in this r-clique.  $\square$

The following r-update rules form an r-clique with a maximal set of predicates  $\{P_1, \dots, P_p\}$ , where each  $P_i$  is a distinguished predicate in each r-update rule in the r-clique:

$r_1$ :  $P_1(\overline{u_{1,1}}), P_2(\overline{u_{1,2}}), \dots, P_p(\overline{u_{1,p}}), A_{1,1}, \dots, A_{1,a_1}, \neg B_{1,1}, \dots, \neg B_{1,b_1} \rightarrow \text{retract } P_1(\overline{u_{1,1}})$

$r_2$ :  $P_1(\overline{u_{2,1}}), P_2(\overline{u_{2,2}}), \dots, P_p(\overline{u_{2,p}}), A_{2,1}, \dots, A_{2,a_2}, \neg B_{2,1}, \dots, \neg B_{2,b_2} \rightarrow \text{retract } P_2(\overline{u_{2,2}})$

:

$r_p$ :  $P_1(\overline{u_{p,1}}), P_2(\overline{u_{p,2}}), \dots, P_p(\overline{u_{p,p}}), A_{p,1}, \dots, A_{p,a_p}, \neg B_{p,1}, \dots, \neg B_{p,b_p} \rightarrow \text{retract } P_p(\overline{u_{p,p}})$

Each  $\overline{u_{ij}}$  is a vector of terms (variables and constants), and for each clique, all terms  $\overline{u_{1,i}} \dots \overline{u_{p,i}}$ , associated with each predicate  $P_i$  in the maximal set must be unifiable.

We relax the original condition (3) of the strict-stratification criterion which was as follows:

- (3) For every distinguished predicate occurring in a positive literal in a rule in  $UR_i$ , all r-update rules relevant to this predicate must be contained within  $\bigcup_{j < i} UR_j$ .

We relax this condition to allow rules such as in the previous example to be legal update rule programs. Condition 3 is modified as follows:

- (3\*) For every distinguished predicate occurring in a positive literal in an r-update rule  $p$ , in  $UR_i$ , any relevant r-update rule  $q$  must be contained within  $\bigcup_{j < i} UR_j$

*unless* the rules  $p$  and  $q$  participate in an r-clique in the partition  $UR_i$ . Then  $p$  and  $q$  may be contained within the same partition  $UR_i$ .

### Definition

2 (or more) r-cliques *overlap* if the predicates in the set of the maximal set of predicates of each clique is not

disjoint and the corresponding vectors of terms are unifiable.  $\square$

### 3.3. Second Relaxation

A similar situation may occur with rules that are used to infer new information. The body of each rule queries some subset of predicates. If the rules (which infer different information), query the same subset of predicates or some mutually overlapping subsets, then it is possible that the stratification conditions will be violated. Sometimes, if one rule executes and asserts some new information, then this prevents the execution of another rule. The relaxation applies to a-update rules. They are allowed to co-exist in the same partition when they have a negative dependency on each predicate in some set.

For this example, we consider the same relations as in Example 3. The rules are based on the knowledge that each person must be either a male or a female. This knowledge is again subject to the constraint that a person cannot be a male and a female simultaneously. As before, the choice predicate is used to select whether the person should be a male or a female. Rule  $r_1$  asserts the fact that the person is a female, but it first queries the database to make sure the person is not a male as well. Similarly, rule  $r_2$  asserts that a person is a male, after having first verified that this person is not a female as well. The rules are as follows:

#### Example 5

$r_1$ :  $\text{person}(x) \neg \text{male}(x), \text{choice}(x, "f") \rightarrow \text{assert female}(x)$

$r_2$ :  $\text{person}(x) \neg \text{female}(x), \text{choice}(x, "m") \rightarrow \text{assert male}(x)$

Again, if we apply the strict stratification criterion, condition (4) requires that  $r_1$  be in a lower partition than  $r_2$  and vice versa. Our solution is to relax the criteria for stratification.

#### Definition

A set of a-update rules participate in an *a-clique* if there is some *maximal* set of predicates  $\{P_1, \dots, P_n\}$  such that each rule in the a-clique is relevant to some predicate from this set; if there is at least one rule in the a-clique which is relevant to each predicate in this set; and if all of the predicates in the set (except the predicate in the head of each rule) are distinguished and occur in a negative literal in each rule of this a-clique.  $\square$

The following a-update rules form an a-clique with a maximal set of predicates  $\{P_1, \dots, P_p\}$ , where each  $P_i$  is a distinguished predicate in each a-update rule in the a-clique, except in the rule relevant to  $P_i$ :

$r_1$ :  $\neg P_2(\overline{u_{1,2}}), \neg P_3(\overline{u_{1,3}}), \dots, \neg P_p(\overline{u_{1,p}}), A_{1,1}, \dots, A_{1,a1}, \neg B_{1,1}, \dots, \neg B_{1,b1} \rightarrow \text{assert } P_1(\overline{u_{1,1}})$

$r_2$ :  $\neg P_1(\overline{u_{2,1}}), \neg P_3(\overline{u_{2,3}}), \dots, \neg P_p(\overline{u_{2,p}}), A_{2,1}, \dots, A_{2,a2}, \neg B_{2,1}, \dots, \neg B_{2,b2} \rightarrow \text{assert } P_2(\overline{u_{2,2}})$

:

$r_p$ :  $\neg P_1(\overline{u_{p,1}}), \neg P_2(\overline{u_{p,2}}), \dots, \neg P_{p-1}(\overline{u_{p,p-1}}), A_{p,1}, \dots, A_{p,ap}, \neg B_{p,1}, \dots, \neg B_{p,bp} \rightarrow \text{assert } P_p(\overline{u_{p,p}})$

As before, we must ensure that the execution of the rule terminates and produces a minimal database. We relax condition (4) of the stratification criterion, which was stated as follows:

- (4) For every distinguished predicate occurring in a negative literal in a rule in  $UR_i$ , all a-update rules relevant to this distinguished predicate must be contained within  $\bigcup_{j < i} UR_j$ .

We modify this condition, as follows:

- (4\*) For every distinguished predicate occurring in a negative literal in an a-update rule  $p$ , in  $UR_i$ , any relevant a-update rule  $q$  must be contained within  $\bigcup_{j < i} UR_j$

unless the rules  $p$  and  $q$  are a-update rules that participate in an a-clique in the partition  $UR_i$ . Then,  $p$  and  $q$  may be contained within the same partition  $UR_i$ .

### Definition

2 (or more) a-cliques *overlap* if the predicates in the set of the maximal set of predicates of each clique is not disjoint and the corresponding vectors of terms are unifiable.  $\square$

We note from the definition for relaxing the stratification conditions that the same predicate may not occur in the maximal set of both an a-clique and an r-clique. However, the definition will allow programs with overlapping cliques. Thus, a predicate may occur in the maximal sets of several a-cliques, or of several r-cliques, simultaneously. The relaxed criterion admits many programs which would have been rejected by the strict stratification criterion.

## 4. A Fixpoint Semantics for Update Rule Programs

We define a fixpoint semantics for acceptable update rule programs which guarantees that the execution of the program terminates and each execution schedule produces a minimal database. Non-terminating behavior results when rules can indefinitely keep asserting and retracting the same tuples. A solution is to explicitly keep track of tuples which are asserted and retracted. Once a tuple is retracted it is not affected by further assertions (of the same tuple), so so that this process cannot occur indefinitely. The execution of update rules is defined based on a monotonic operator  $T_{UR}$ . A fixpoint is non-deterministically obtained for the update rule program  $UR$ . The monotonic operator  $T_{UR}$  is non-deterministically applied to all the rules in each partition,  $UR_i$ , in the order  $i = 1, 2, \dots, n$ . A fixpoint  $EDB_i$  is computed for each partition. Each  $EDB_i$  is represented by the corresponding relations, after the execution of all the rules in the partition that can be executed. Thus, a sequence of fixpoints  $EDB_1, EDB_2, \dots$ , corresponding to each partition  $UR_1, UR_2, \dots$ , are computed, in turn. The final updated database is  $EDB_n$  and will contain all the answers obtained from the update rule program.

Consider a domain,  $D_{UR}$ , for an update rule program to be the set of all possible ground atoms, i.e., atoms with no variables, of the form **assert** $P(\bar{t})$  and **retract** $P(\bar{t})$ , where  $P$  is any predicate and  $\bar{t}$  is a vector of constants from the ordered set  $C$ .

### Definition

$EDB_0 = \{\mathbf{assert}P(\bar{t}) \mid P(\bar{t}) \in EDB_{init}\}$ .  $\square$

$EDB_0$  is a set of tuples in the relations **assert** $P$ , **assert** $Q$ , etc., corresponding to the facts of the initial database  $EDB_{init}$  and the relations **retract** $P$ , **retract** $Q$ , etc. will initially be empty, since the database only contains facts, initially.

### Definition

Given a subset  $S$  of  $D_{UR}$ , we have the following:

- $S$  entails  $P(\bar{t})$  if **assert** $P(\bar{t}) \in S$  and **retract** $P(\bar{t}) \notin S$ .
- $S$  entails  $\neg P(\bar{t})$  if and only if  $S$  does not entail  $P(\bar{t})$ . This occurs with the following:

**assert** $P(\bar{t}) \in S$  and **retract** $P(\bar{t}) \in S$  or

**assert** $P(\bar{t}) \notin S$  and **retract** $P(\bar{t}) \notin S$ .

- The following cannot occur in  $S$ : **assert** $P(\bar{t}) \notin S$  and **retract** $P(\bar{t}) \in S$ ; this is due to the syntactic restriction on the language.  $\square$

### Definition

Given the ordered set of constants  $C$  over which  $D_{UR}$  is defined, we have  $\text{eval}(\text{op}_i, \text{arg}_{i,1}, \text{arg}_{i,2})$  is true if

- (1)  $op_i$  is in the set  $\{ \leq, <, \geq, >, =, \neq \}$ ; and
- (2) each of the arguments  $arg_{i,1}$  and  $arg_{i,2}$  are constants in the set  $C$ , or they are range restricted variables occurring in a positive literal, i.e., they can be instantiated to constants in the set  $C$ ; and
- (3)  $(arg_{i,1} op arg_{i,2})$  when interpreted in the usual fashion evaluates to true.

Otherwise,  $eval(op_i, arg_{i,1}, arg_{i,2})$  evaluates to false. □

We define an operator  $T_{UR}$  which, when applied to a rule in a partition of the update rule program, will update the EDB relations. Since  $T_{UR}$  is applied non-deterministically, we first define an intermediate operator  $M_{UR}$ .

#### Definition

Let  $M_{UR} : 2^{D_{UR}} \rightarrow 2^{D_{UR}}$  be a mapping from subsets of  $D_{UR}$  to subsets of  $D_{UR}$ , defined as follows, where  $S$  is a subset of  $D_{UR}$ :

$M_{UR}(S) = \{ \mathbf{assertP}(\bar{t}) \mid \text{there is a ground instance of a rule } p \text{ (i.e., an instance of } p \text{ where all the variables are substituted by constants from } C \text{):}$

$A_1, A_2, \dots, A_n, \neg B_1, \neg B_2, \dots, \neg B_m, eval(op_1, arg_{1,1}, arg_{1,2}), eval(op_2, arg_{2,1}, arg_{2,2}), \dots, eval(op_k, arg_{k,1}, arg_{k,2}),$   
 $\rightarrow \mathbf{assertP}(\bar{t})$

of an update rule such that  $S$  entails  $B_1, B_2, \dots, B_n, \neg C_1, \neg C_2, \dots, \neg C_m$ ,

and each  $eval(op_i, arg_{i,1}, arg_{i,2})$  evaluates to true }.

$\cup$

$\{ \mathbf{retractP}(\bar{t}) \mid \text{there is a ground instance } p:$

$A_1, A_2, \dots, A_n, \neg B_1, \neg B_2, \dots, \neg B_m, eval(op_1, arg_{1,1}, arg_{1,2}), eval(op_2, arg_{2,1}, arg_{2,2}), \dots, eval(op_k, arg_{k,1}, arg_{k,2}),$   
 $\rightarrow \mathbf{retractP}(\bar{t})$

of an update rule such that  $S$  entails  $B_1, B_2, \dots, B_n, \neg C_1, \neg C_2, \dots, \neg C_m$ ,

and each  $eval(op_i, arg_{i,1}, arg_{i,2})$  evaluates to true }.

□

#### Definition

Let  $T_{UR} : 2^{D_{UR}} \rightarrow 2^{D_{UR}}$  be a mapping from a subset of  $D_{UR}$  to a subset of  $D_{UR}$  defined as follows:

$T_{UR}(S) = S \cup \{Q\}$  where  $Q \in M_{UR}(S)$  and  $Q \notin S$ . □

Note that  $Q$  is selected non-deterministically.

### 4.1. Computing the Operational Fixpoint

Each  $EDB_i$  is computed iteratively, applying the operator  $T_{UR}$  to the rules in each partition  $UR_i$  until a fixpoint  $EDB_i$  is reached.  $T_{UR}$  will select a single rule for execution in each step and this will result in a possibly non-deterministic execution. Since  $T_{UR}$  is monotonic, a fixpoint is obtained for each partition,  $UR_i$ , when there are no longer any rule that can change  $EDB_i$ . Processing for the update rule program terminates when the fixpoint for rules in  $UR_n$  is reached and  $EDB_n$  is computed.

$$EDB_1 = T_{UR_1} \uparrow \omega (EDB_0)$$

$$EDB_2 = T_{UR_2} \uparrow \omega (EDB_1)$$

$\vdots$

$$EDB_n = T_{UR_n} \uparrow \omega (EDB_{n-1})$$

**Theorem**

$EDB_i = T_{UR} \uparrow \omega (EDB_{i-1})$  is a fixpoint of  $T_{UR}$ , for  $i=1,2,\dots,n$ .  $EDB_n$  is a fixpoint for  $T_{UR}$ .

We observe from the definition of the operator  $T_{UR}$  that it is growing monotonically, i.e.,  $S \subseteq T_{UR}(S)$  and that  $T_{UR} \uparrow n (S) \subseteq T_{UR} (T_{UR} \uparrow n (S))$ . Since the domain  $D_{UR}$  for the update rule program is finite, after some finite number of applications  $n$ ,  $n < \omega$ , the operator will not add any new element of the form (**assert**  $P(\bar{t})$ ) or (**retract**  $P(\bar{t})$ ) to  $EDB_i$ .

Thus,  $EDB_i = T_{UR} \uparrow \omega (EDB_{i-1})$  is a fixpoint of  $T_{UR}$  for  $i = 1, 2, \dots, n$ .  $\square$

We note that there may be several ways to partition a program, and to identify overlapping cliques (within the same partition). However, this does not have an effect on computing the fixpoint.

**4.2. Some Relevant Properties of the Update Rule Programs**

We present some relevant properties of the update rule programs, UR. These properties, represented by the following lemmas, 4.1 through 4.6, follow from the criterion for strict stratification or the relaxed criterion presented in Section 3. The proofs are straightforward and are omitted for readability. The results will be applied to the proof of the equivalence between the fixpoint semantics for UR programs and the declarative semantics, which is discussed in the next few sections. The following properties (expressed in the propositional case) hold true for the UR programs:

**Lemma 4.1**

Let  $UR = UR_1 \cup \dots \cup UR_n$  be a UR program. Let  $UR^- = UR_1^- \cup \dots \cup UR_n^-$  be the resulting program after eliminating all the a-update rules in UR which participate in each a-clique and all the r-update rules which participate in each r-clique, in all the partitions of UR. Then,  $UR^-$  is a strictly-stratified UR program.  $\square$

**Lemma 4.2**

Let UR be an update rule program. Then the set of all atoms that occur in any maximal set associated with any a-clique and the set of all atoms that occur in any maximal set associated with any r-clique are mutually disjoint.  $\square$

**Lemma 4.3**

Let  $UR = UR_1 \cup \dots \cup UR_n$  be an update rule program. Let the retractable atom  $P_i$  occur in the maximal set associated with some r-clique in the partition  $UR_j$ . Then,  $P_i$  may not be distinguished in a rule in any partition  $\bigcup_{k < j} UR_k$ . Further, all rules  $q$  in  $UR_j$  in which  $P_i$  is distinguished must be r-update rules and they must participate in this r-clique.  $\square$

**Lemma 4.4**

Let  $UR = UR_1 \cup \dots \cup UR_n$  be an update rule program. Let the atom  $P_i$  occur in the maximal set associated with some a-clique in the partition  $UR_j$ . Then,  $P_i$  may not be distinguished in a rule in any partition  $\bigcup_{k < j} UR_k$ . Further, all rules  $q$  in  $UR_j$  in which this  $P_i$  is distinguished (and occurs in a negative literal) must be an a-update rule and they should participate in the a-clique.  $P_i$  may also be a distinguished atom (occurring in a positive literal) in other rules in  $UR_j$ . Finally,  $P_i$  is unretractable in the UR program.  $\square$

**Lemma 4.5**

Let  $p$  and  $q$  be example a-update rules or r-update rules in  $UR_i$ . Then, all distinguished atoms in  $p$  or  $q$ , which do not occur in a maximal set of atoms associated with some r-clique in  $UR_i$ , must be unretractable in  $UR_i$ .  $\square$

The proofs of these Lemmas follow directly from the relaxed stratification criterion for the update rule programs.



#### Lemma 4.6

For any set of a-update rules participating in an a-clique in partition  $UR_k$ , or for any set of r-update rules participating in an r-clique, at most one of the rules in the a-clique (or one of the rules in the r-clique) may execute (for a given assignment for the variables, in the first order case).

The proof for the a-clique follows from the fact that for each a-update rule relevant to  $P_i$ , participating in an a-clique, all  $P_j$ ,  $j \neq i$ , in the maximal set, must be distinguished and occur in a negative literal in the a-update rule. Thus, for the execution of any a-update rule in the a-clique, relevant to  $P_i$ , none of the  $P_j$ ,  $j \neq i$ , must be entailed by  $EDB_k$ . The execution of any one of the a-update rules in the a-clique will add some **assert**  $P_i$ , to  $EDB_k$  and  $EDB_k$  will entail this  $P_i$ . From the definition of the stratification criterion, this  $P_i$  is unretractable in this partition, It follows that since this  $P_i$  is distinguished and occurs in a negative literal in each of the other a-update rules in the a-clique, none of these other rules may be executed. Thus, at most one a-update rule in an a-clique may be executed.

Similarly, for each r-update rule participating in an r-clique, all  $Q_j$  in the maximal set must be distinguished atoms in these r-update rules. Thus, for the execution of any r-update rule in the r-clique, each of the  $Q_j$  must be entailed by  $EDB_k$ . The execution of any one of the r-update rules will add some **retract**  $Q_j$ , to  $EDB_k$  and  $EDB_k$  will not entail this  $Q_j$ . It follows that none of the other rules may be executed and at most one r-update rule in an r-clique will execute.  $\square$

### 4.3. Set-oriented and Tuple-oriented Execution of Rules

The difference between a tuple-oriented and a set-oriented execution, when the action(s) in the head of the rules update one tuple, or a set of tuples, respectively, can be informally stated, as follows:

When the literals in the body of an update rule are evaluated against the relations of the database, and when a single tuple or a set of tuples can be retrieved, corresponding to the action(s) in the head of the update rule, do the following execution options produce identical fixpoints?

- (1) a *tuple-oriented execution* -- execute the update action(s) in the head of the rule, corresponding to a single selected tuple, and then re-evaluate the literals in the body of the rule against the updated database.
- (2) a *set-oriented execution* -- execute the update action(s) in the head of the rule, corresponding to a single tuple, some subset of the set of tuples or all the tuples in the set, simultaneously.

In our case, the syntactic criterion for admitting UR programs ensures that the fixpoint(s) obtained from the set-oriented execution are a subset of the set of fixpoints corresponding to a tuple-oriented execution. This may be of practical significance since the set-oriented execution is more efficient. We do not discuss this issue in detail in this paper, and refer the reader to a discussion of implementation issues in [PaRa94].

## 5. A Declarative Semantics for Update Rule Programs

We have described a fixpoint semantics for UR programs which is guaranteed to terminate and to produce a set of fixpoints for UR. To obtain a declarative semantics, we first associate a normal logic program  $\overline{UR}$  for a given UR program. We use the *stable model* semantics of Gelfond and Lifschitz [GeLi88] to obtain a set of minimal models for  $\overline{UR}$  which are called *stable models*. We show the equivalence between the set of fixpoints for UR and the set of stable models for  $\overline{UR}$ .

## 5.1. Obtaining a Normal Logic Program

To obtain a normal logic program  $\overline{UR}$ , the a-update rules that assert new tuples are treated like deductive rules which define the asserted tuple. The r-update rules are similar to *denial* integrity constraints proposed in [KoSa89, KoSa90]. A *denial* integrity constraint is given by a denial which is a conjunction of literals, followed by a *retractable* atom which is enclosed within [ ]. The retractable atom must occur in the conjunction of literals. An example is as follows:

$$L_1, L_2, \dots, L_n \rightarrow. [L_i].$$

The meaning associated with this constraint is that  $L_1 \wedge L_2 \wedge \dots \wedge L_n$  cannot be true in the database, if the database is to be consistent with the denial. If the denial is violated in the database, then consistency can be restored by making one of the positive literals in the denial false. A single atom  $L_i$  which occurs in the denial is chosen as the retractable atom, and is specified as such in the constraint.

If we examine the following r-update rule:

$$r_i: P, A_1, \dots, A_n, \neg B_1, \dots, \neg B_m \rightarrow \mathbf{retract} P$$

since the atom  $P$  which is retracted by the r-update rule must occur in the conjunction in the body of the rule, the r-update rule is syntactically similar to the denial integrity constraints. We consider the conjunction in the body of the r-update rule to correspond to the denial.

If we consider the forward chaining semantics of update rule programs, the r-update rules query the database to determine if the condition in the body of the rule is true, i.e., the database must entail  $P$  and every literal  $A_k$ , but must not entail any literal  $B_k$ . The condition will include the fact,  $P$ , that is to be deleted. When the condition is true, the corresponding fact  $P$ , (which must exist), is deleted, so that the condition is no longer true in the database. Operationally, this is very close to the task of maintaining an integrity constraint. When we consider related research, we will examine different semantics where the deletion action corresponds to a negation.

There has been considerable research on the problem of maintaining consistency in databases, with respect to a set of integrity constraints. Informally, a database must satisfy its integrity constraints as it changes over time. Usually, an update to the database (more precisely an update to facts in the database) may cause the violation of an integrity constraint. Such updates must be rejected or modified. Sometimes, the database itself, i.e., the initial facts and the facts that are inferred from the rules, may be inconsistent with the constraints, and the database must be modified to maintain consistency with the constraints. This is the approach we have taken to provide a semantics for the updates.

The translation to obtain  $\overline{UR}$  uses as input the update rules of the program  $UR$  and the initial database  $EDB_{init}$ , and produces the logic program  $\overline{UR}$  as output. For simplicity, we show this transformation for the propositional case. The translation in the first order case is included in Appendix 1.

### A Translation to obtain a Normal Logic Program

#### Step 1.

Each r-update rule which is relevant to the atom  $P$  transforms each a-update rule in  $UR$  which is relevant to the atom  $P$ , or the fact  $P$  if it occurs in  $EDB_{init}$ .

Suppose the a-update rule relevant to  $P$ , is of the following form:

$$A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow \mathbf{assert} P.$$

where  $a$  or  $b$  could be equal to 0, i.e.,  $P$  is a fact, is transformed by each r-update rule relevant to  $P$ , which is as follows:

$P, C_1, \dots, C_c, \neg D_1, \dots, \neg D_d \rightarrow \mathbf{retract} P.$

The following rules are placed in  $\overline{UR}$ , where  $P^*$  is a special unused atom associated with each atom  $P$  in the program:

$C_1, \dots, C_c, \neg D_1, \dots, \neg D_d, \rightarrow P^*.$

$\neg P^*, A_1, \dots, A_a, \neg B_1, \dots, \neg B_b, \rightarrow P.$

### Step 2

Each fact, say  $P$ , in  $EDB_{init}$  which is not modified by any r-update rule, i.e., there are no r-update rules in  $UR$  relevant to  $P$  is placed in  $\overline{UR}$ .

Each a-update rule relevant to the atom  $P$  which is as follows:

$A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow \mathbf{assert} P.$

and which is not modified by any r-update rule in  $UR$  will derive the following rule in  $\overline{UR}$ :

$A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow P.$

## 5.2. Stable Model Semantics

The program  $\overline{UR}$  may not be a stratified logic program and may not have a unique minimal model. We use the stable model semantics to obtain a meaning for this program. A model theory for normal logic programs [GeLi88] characterizes the meaning of a normal logic program by a set of minimal models called *stable models*, which are defined using the Gelfond-Lifschitz transformation. This transformation is defined as follows.

### Definition

Let  $P$  be a normal logic program and let  $I$  be an interpretation.

$P^I = \{ A \leftarrow B_1, \dots, B_n : A \leftarrow B_1, \dots, B_n, \neg D_1, \dots, \neg D_m \text{ is a ground instance of a clause in } P \text{ and } \{D_1, \dots, D_m\} \cap I = \Phi \}$

$P^I$  is the Gelfond-Lifschitz transformation of  $P$  with respect to  $I$ , where the  $A_i, B_j$  and  $D_l$  are atomic formulae.  $\square$

The result of the Gelfond-Lifschitz transformation is a negation-free (definite) program. Stable models for non-disjunctive logic programs may now be defined as follows:

### Definition

Let  $P$  be a definite normal program.  $M$  is a stable model of  $P$  iff  $M$  is the unique minimal model of  $P^M$ .  $\square$

## 5.3. Equivalence of the Fixpoint and Declarative Semantics

### Theorem

Let  $\overline{UR}$  be the normal logic program derived from the update rule program  $UR$  corresponding to the set of update rules and the initial database  $EDB_{init}$ .

A fixpoint of the operational semantics for  $UR$  represented by a final updated database  $EDB_n$ , is identical to one of the (k) stable models for the normal logic program  $\overline{UR}$ , as characterized by the stable model semantics.

All of the (k) stable models for the corresponding normal logic program  $\overline{UR}$ , as characterized by the stable model semantics, can each be non-deterministically obtained as a final updated database  $EDB_n$ , from the UR program.

The following figure is a graphical representation of this equivalence.

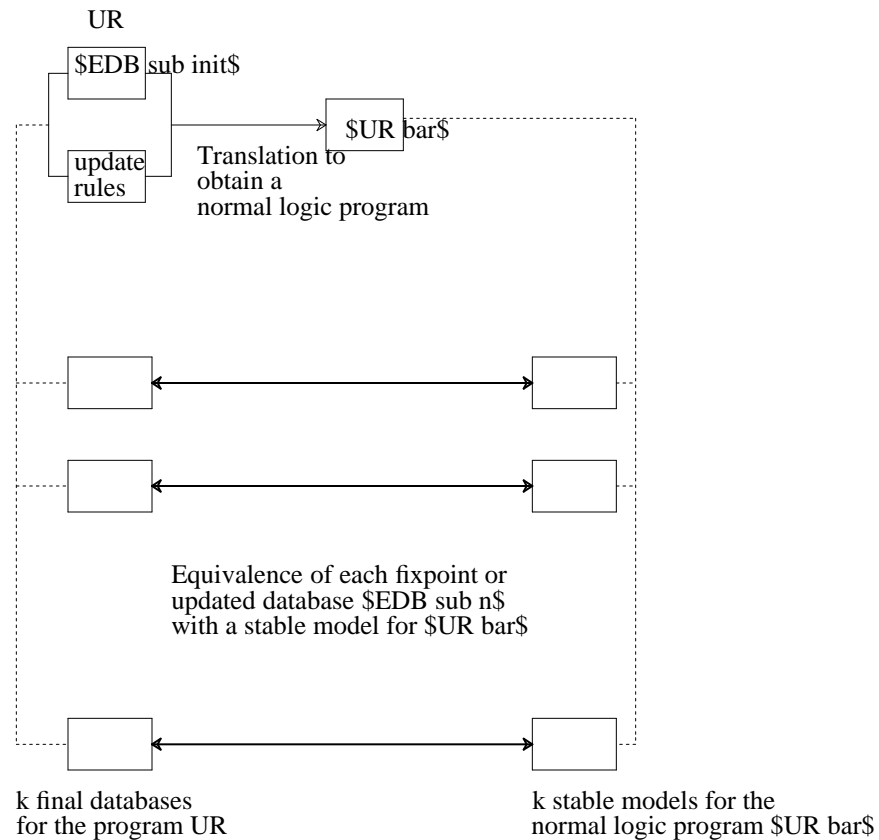


Figure 1. Equivalence Between the Fixpoint and Declarative Semantics

## 6. Proving the Equivalence of the Fixpoint and Declarative Semantics

In this section, we present the proof of equivalence of the fixpoint and declarative semantics.

### 6.1. Some Results from the Stable Model Semantics for Normal Logic Programs

We first present some results on the stable model semantics and the corresponding minimal (stable) models, for some normal logic programs  $\overline{UR}$ . The normal logic programs that we consider will correspond to some interesting UR program fragments. These results will be applied to the proof of equivalence of the fixpoint and declarative semantics. The results are represented by the following lemmas, (expressed in the propositional case):

#### Lemma 6.1

Let  $\overline{UR}$  be a normal logic program derived from an update rule program fragment comprising a single (non-overlapping) r-clique with a maximal set of atoms  $\{P_1, \dots, P_p\}$ , where each  $P_i$  is a retractable atom, and the update rules relevant to each  $P_i$ . Let  $P_i^*$  be the special atoms associated with each atom  $P_i$  in the maximal set.

Suppose that all other distinguished literals in the rules may be ignored. Then there are  $p$  possible stable models for  $\overline{UR}$ . In each stable model, exactly one of the  $p$  atoms in the maximal set will be false, the corresponding special atom will be true, all other atoms in the maximal set will be true, and the corresponding special atoms will be false.

To prove this result, let  $\overline{UR}$  be derived from a program fragment comprising a set of  $r$ -update rules  $s_1, \dots, s_p$ , which form an  $r$ -clique with a maximal set of atoms  $\{P_1, \dots, P_p\}$ . To obtain  $\overline{UR}$ , we also consider any  $a$ -update rules  $q_k$  relevant to some atom  $P_k$  occurring in the maximal set of this  $r$ -clique. For simplicity, we assume that there are exactly  $p$   $r$ -update rules and  $p$   $a$ -update rules, each one relevant to exactly one  $P_i$ . However, this lemma also applies to the case where there may be several rules relevant to each atom. Since the  $r$ -clique is non-overlapping, none of the  $r$ -update rules  $s_k$  participate in another  $r$ -clique. For notational convenience, all other distinguished literals in the rules are represented using some conjunction  $Conj-i,j$ .

The normal logic program  $\overline{UR}$  from the rules  $p_k$  and  $s_k$ , is as follows:

$$\begin{array}{lll}
 r_{1,1}: & P_2, \dots, P_p, Conj-1,1 \rightarrow P_1^* & \\
 : & & \text{each } r_{1,k} \text{ is from } s_k: \quad P_1, \dots, P_k, \dots, P_p, Conj-1,k \rightarrow \mathbf{retract } P_k \\
 r_{1,p}: & P_1, \dots, P_{p-1}, Conj-1,p \rightarrow P_p^* & \\
 \\
 r_{2,1}: & \neg P_1^*, Conj-2,1 \rightarrow P_1 & \text{each } r_{2,k} \text{ is from } s_k \text{ transforming } q_k: \quad Conj-2,k \rightarrow \mathbf{assert } P_k \\
 : & & \\
 r_{2,p}: & \neg P_p^*, Conj-2,p \rightarrow P_p & 
 \end{array}$$

Assume that for all the stable models, all the distinguished atoms in each conjunction  $Conj-i,j$  are true and each distinguished literal that occurs negatively in each conjunction is false. We may ignore each of these conjunctions in the rules. This assumption will be clarified later.

Now consider some model  $M_k$  in which one  $P_k^*$  and all  $P_t$ ,  $t \neq k$ , (from the maximal set of atoms) are true.  $P_k$  and all  $P_t^*$ ,  $t \neq k$ , are false in this model. To prove this is a stable model, we first apply the Gelfond-Lifschitz transformation to obtain the negation-free program  $\overline{UR}_{M_k}$ . All rules  $r_{1,1}$  through  $r_{1,p}$  will be unchanged in  $\overline{UR}_{M_k}$  (ignoring the conjunctions). All rules  $r_{2,1}$  through  $r_{2,p}$ , except  $r_{2,k}$ , will be modified in  $\overline{UR}_{M_k}$  so that the special distinguished literals  $P_t^*$ ,  $t \neq k$ , occurring negatively in these rules are eliminated. These rules will be as follows:

$$r_{2,t} : Conj-2,t \rightarrow P_t$$

The rule  $r_{2,k}$  (or set of rules) will be eliminated in  $\overline{UR}_{M_k}$ .

The above transformation is independent of whether there are one or more  $a$ -update rules and  $r$ -update rules relevant to each  $P_i$  and the program that is obtained with multiple rules will be similar. Now, from the rule (rules)  $r_{1,k}$  (ignoring the conjunctions), we have  $P_k^*$  must be true. From each rule (rules)  $r_{2,t}$ ,  $t \neq k$ , we have each  $P_t$ ,  $t \neq k$ , must be true. Further, since rule (rules)  $r_{2,k}$  is eliminated,  $P_k$  must be false. Since  $P_k$  is a distinguished atom in each of the rules  $r_{1,t}$ ,  $t \neq k$ , each  $P_t^*$ ,  $t \neq k$ , must be false. Thus, we have  $M_k$  is a stable model for the program.

Next, we prove that each such  $M_k$  is a minimal model for  $\overline{UR}_{M_k}$ . We prove this by considering the following two cases:

### Case 1

Suppose there is a model  $M_{c1}$  which is a proper subset of  $M_k$ , in which  $P_k$  and one (or more)  $P_x$  from the maximal set are false. However, from each rule (rules)  $r_{2,x}$  in  $\overline{UR}_{M_k}$ , we have each  $P_x$ ,  $x \neq k$  must be true. Thus, this model is not stable.

### Case 2

Suppose there is a model  $M_{c2}$  which is a proper subset of  $M_k$ , in which the special atom  $P_k^*$  is false, i.e., none

of the special atoms are true. Then, from rule (rules)  $r_{1,k}$  in  $\overline{UR}_{M_k}$ , since  $P_k^*$  is false, at least one  $P_x$ ,  $x \neq k$ , must be false. However, from the rule (rules)  $r_{2,x}$  in  $\overline{UR}_{M_k}$ , we have each  $P_x$ ,  $x \neq k$  must be true. Thus, this model is not stable.

Finally, we prove that the models  $M_k$  enumerated in Lemma 6.1, are the only stable models for the program, by considering these following two cases:

### Case 3

Suppose there is a model  $M_{c3}$  in which all atoms in the maximal set are true. But we know that the rule (rules)  $r_{2,k}$ , relevant to  $P_k$  is eliminated in  $\overline{UR}_{M_k}$ . Thus, this model is not a stable model.

### Case 4

Suppose there is a model  $M_{c4}$  in which some  $P_k^*$  and one (or more)  $P_y^*$ ,  $y \neq k$ , is true. From rule (rules)  $r_{1,y}$  in  $\overline{UR}_{M_k}$ , it follows that  $P_k$  must be true. However, the rule (rules)  $r_{2,k}$ , relevant to  $P_k$ , is eliminated in  $\overline{UR}_{M_k}$ . Thus, this model is not a stable model.

Thus, we have shown that there are  $p$  possible stable models, independent of whether there are several rules relevant to each  $P_i$  in the maximal set.  $\square$

### Lemma 6.2

Let  $\overline{UR}$  be a normal logic program derived from an update rule program fragment comprising overlapping  $r$ -cliques with maximal sets of atoms  $\{P_1, \dots, P_p, C_1, \dots, C_c\}$  and  $\{Q_1, \dots, Q_q, C_1, \dots, C_c\}$ , where  $\{C_1, \dots, C_c\}$  are common to both maximal sets. We also consider the relevant  $a$ -update rules. Each  $P_i$ ,  $Q_i$  and  $C_i$  is a retractable atom. Let  $P_i^*$ ,  $Q_i^*$  and  $C_i^*$  be the special atoms associated with the atoms in the maximal sets. Assume that all other distinguished literals in the rules in  $\overline{UR}$  may be ignored. Then there will be  $(p * q + c)$  possible stable models. In  $(p * q)$  of these models, exactly one  $P_{k1}$  and exactly one  $Q_{k2}$ , will be false and exactly one  $P_{k1}^*$  and exactly one  $Q_{k2}^*$  will be true. In addition all other  $P_t$ ,  $t \neq k1$ , and all other  $Q_t$ ,  $t \neq k2$ , will be true and these  $P_t^*$  and  $Q_t^*$ , will be false. Finally, all  $C_s$  will be true and all  $C_s^*$  will be false. In  $c$  of these extensions, exactly one atom  $C_{k3}$  which is common to both the maximal sets is false,  $C_{k3}^*$  is true, all other atoms  $C_t$ ,  $t \neq k3$ , will be true and all  $C_t^*$ ,  $t \neq k3$ , will be false. In addition, in all these extensions, all  $P_s$  and all  $Q_r$  will also be true and all  $P_s^*$  and all  $Q_r^*$  will be false.

The proof of Lemma 6.2 is omitted.  $\square$

### Lemma 6.3

Let  $\overline{UR}$  be a normal logic program derived from an update rule program fragment comprising a single non-overlapping  $a$ -clique with a maximal set of atoms  $\{P_1, \dots, P_p\}$ . Assume that all other distinguished literals in the rules in  $\overline{UR}$  may be ignored. Then there are  $p$  possible stable models for  $\overline{UR}$ . In each stable model, exactly one of the  $p$  atoms in the maximal set will be true and all other atoms will be false.

To prove this result, let  $\overline{UR}$  be derived from a program fragment comprising a set of  $a$ -update rules  $s_1, \dots, s_p$ , which form an  $a$ -clique with the maximal set of atoms  $\{P_1, \dots, P_p\}$ . For simplicity, we assume that there are exactly  $p$  such rules, each one relevant to exactly one  $P_k$ . However, the lemma also applies with several rules relevant to each  $P_k$ . Since the  $a$ -clique is non-overlapping, none of the  $a$ -update rules  $s_k$  participate in another  $a$ -clique. All other distinguished literals in the rules are represented using the conjunctions  $Conj-k$ . The normal logic program  $\overline{UR}$  obtained from the rules is as follows:

$$r_1 : \neg P_2, \dots, \neg P_p, Conj-1 \rightarrow P_1$$

$$r_k \text{ is from } s_k : \neg P_1, \dots, \neg P_{k-1}, \neg P_{k+1}, \dots, \neg P_p \rightarrow \text{assert } P_k$$

$$r_p : \neg P_1, \dots, \neg P_{p-1}, Conj-p \rightarrow P_p.$$

Assume that for all the stable models, all the distinguished atoms in each conjunction are true and each distinguished literal that occurs negatively in each conjunction is false. We ignore each of these conjunctions in the



rules.

Now consider some model  $M_k$  in which one  $P_k$  is true and all  $P_t$ ,  $t \neq k$ , (from the maximal set) are false. To prove this is a stable model, we first apply the Gelfond-Lifschitz transformation to obtain the negation free program  $\overline{UR}_{M_k}$ . All rules  $r_1$  through  $r_p$  except the rule  $r_k$  will be eliminated in  $\overline{UR}_{M_k}$  since  $P_k$  is a distinguished literal occurring negatively in these rules. The rule  $r_k$  will be modified as follows:

$$r_k : \text{Conj-}k \rightarrow P_k$$

The transformed program will be similar if there are several rules relevant to each of the predicates  $P_i$ . From this rule (rules)  $r_k$ , ignoring the conjunction, we have  $P_k$  must be true. Thus each  $M_k$  is a stable model.

To prove that each  $M_k$  is a minimal model, we only need consider the empty interpretation since it is the only proper subset of any  $M_k$ . However, from the rule (or rules)  $r_k$  in  $\overline{UR}_{M_k}$ , we have  $P_k$  is true, and so this is not stable.

There can be no other possible models for  $\overline{UR}_{M_k}$  which are minimal models. Thus, there are  $p$  possible models, and this independent of whether there are multiple rules relevant to each  $P_i$ .  $\square$

**Lemma 6.4:**

Let  $\overline{UR}$  be a normal logic program derived from an update rule program fragment comprising overlapping a-cliques with maximal sets of atoms  $\{P_1, \dots, P_p, C_1, \dots, C_c\}$  and  $\{Q_1, \dots, Q_q, C_1, \dots, C_c\}$ , where  $\{C_1, \dots, C_c\}$  are common to both maximal sets. Assume we ignore all other literals in  $\overline{UR}$ . Then there will be  $(c + p * q)$  possible stable models. In  $(p * q)$  of these models, exactly one  $P_{k1}$  and exactly one  $Q_{k2}$ , will be true. All other  $P_t$ ,  $t \neq k1$ , and all other  $Q_t$ ,  $t \neq k2$ , as well as all other  $C_t$ , will be false. In  $c$  of these models, exactly one atom  $C_{k3}$  which is common to both the maximal sets is true. All other atoms  $C_t$ ,  $t \neq k3$ , will be false as well as all  $P_s$  and all  $Q_r$ .

The proof of Lemma 6.4 is omitted.  $\square$

## 6.2. Soundness and Completeness of the Fixpoint Semantics

The proof that the fixpoint semantics is sound and complete wrt the declarative semantics for the corresponding normal logic program  $\overline{UR}$  is presented in this section. This proof builds upon a previous result for strictly-stratified UR programs. In [Rasc94], we have shown that a stratified logic program can be obtained corresponding to each strictly-stratified UR program. We have further shown that the fixpoint semantics for the strictly-stratified UR program is sound and complete with respect to the declarative semantics for the corresponding stratified logic program. This theorem is as follows:

**Theorem [Rasc94]**

Suppose  $UR^- = UR^-_0 \cup UR^-_1 \cup \dots \cup UR^-_n$  is a strictly-stratified UR program. Let  $\overline{UR}^-(0, \dots, i)$  be the stratified logic program corresponding to the initial database  $EDB_0$  (or  $UR^-_0$ ) and the rules in the partitions,  $UR^-_1 \cup \dots \cup UR^-_i$ . Let  $M_{\overline{UR}^-(0, \dots, i)}$  be the standard model for  $\overline{UR}^-(0, \dots, i)$ . Then,  $EDB^-_i$  entails  $M_{\overline{UR}^-(0, \dots, i)}$ . In other words, if  $P$  is true in  $M_{\overline{UR}^-(0, \dots, i)}$  then  $EDB^-_i$  entails  $P$ , and if  $P$  is false in  $M_{\overline{UR}^-(0, \dots, i)}$  then  $EDB^-_i$  does not entail  $P$ .

Recall from Lemma 4.1 that by removing all the r-update rules that participate in any r-cliques and the a-update rules that participate in any a-cliques, we can obtain a strictly-stratified UR program from the update rule program  $UR$ . Building upon this previous theorem for the strictly-stratified UR programs, we will start from a strictly-stratified UR program and then introduce into each partition in turn, the r-update rules that participate in r-cliques or the a-update rules that participate in a-cliques. We allow these rules to execute until a fixpoint is obtained for this partition. We will then show that the fixpoint semantics for the resulting UR program, after these r-update rules and the a-update rules are introduced into each partition, is sound and complete wrt the declarative semantics

for the corresponding normal logic program  $\overline{UR}$ . We present the proof in the propositional case. It should hold in the function free first order case where all terms are ground.

### Definition

Let  $\overline{UR}$  be a normal logic program. Then, a distinguished atom  $P$  in a rule has a positive dependency on an atom  $Q$  if and only if  $P$  occurs in the head of a rule  $p$  and  $Q$  is a distinguished atom in  $p$ , or if  $R$  is a distinguished atom in the rule  $p$  and  $R$  has a positive dependency on  $Q$ .

### Theorem

Let  $\overline{UR}$  be the normal logic program derived from an update rule program  $UR$ .

Each fixpoint from the  $UR$  program, represented by a final updated database  $EDB_n$ , will entail one of the stable models that characterize  $\overline{UR}$ .

Each of the stable models that characterize  $\overline{UR}$  is entailed by a fixpoint or final updated database  $EDB_n$ , from the  $UR$  program. <sup>1</sup>

### Proof

Let  $UR = UR_0 \cup UR_1 \cup \dots \cup UR_n$  be an update rule program  $UR$ , where  $UR_0 = EDB_0$ .

Let  $\overline{UR}(0, \dots, i)$  be the normal logic program derived from the initial database  $EDB_0$  and the rules in the partitions,  $UR_1 \cup \dots \cup UR_i$ .

Let  $UR^- = UR_0^- \cup UR_1^- \cup \dots \cup UR_n^-$  be the strictly-stratified  $UR$  program that is obtained by eliminating the  $r$ -update rules in each of the  $r$ -cliques and the  $a$ -update rules in each of the  $a$ -cliques, in each of the partitions of  $UR$ .

Let  $\overline{UR}^-(0, \dots, n)$  be the stratified logic program corresponding to the strictly-stratified  $UR$  program  $UR^-$ .

### Base Case

The base case is to prove that  $EDB_0$  entails  $M_{\overline{UR}^-(0)}$ . By definition, there are no update rules in  $UR_0$  and  $\overline{UR}^-(0)$  is identical to the initial database  $EDB_{init}$ . The proof trivially follows from the definition of  $EDB_0$ .

### Inductive Case

Given  $EDB_i$  entails a stable model  $M_i$  for the corresponding normal logic program  $\overline{UR}(0, \dots, i)$ , and

$EDB_{i+1}^- = T_{UR_{i+1}} \uparrow \omega (EDB_i)$ , the fixpoint for the update rules in  $UR_{i+1}^-$  evaluated over  $EDB_i$ , entails the standard model for the stratified logic program derived from  $\overline{M_i} \cup \overline{UR}^-(i+1)$  [Rasc94],

then, each  $EDB_{i+1}$  entails a stable model for the corresponding normal logic program  $\overline{UR}(0, \dots, i+1)$ . In addition, each stable model for  $\overline{UR}(0, \dots, i+1)$  is entailed by some fixpoint  $EDB_{i+1}$  for the update rules in  $UR_{i+1}$ .

Suppose we introduce the update rules in  $UR_{i+1}$  which participate in either  $r$ -cliques or  $a$ -cliques. Executing these update rules will only affect the atoms that occur in some maximal set associated with an  $r$ -clique or an  $a$ -clique. By Lemma 4.2, these sets of atoms are mutually exclusive, so we can independently consider the effect of

<sup>1</sup> The special literals  $P'$ , corresponding to each literal  $P$  in some maximal set only occur in the corresponding normal logic program and they do not occur in the update rule program. They are not entailed in the fixpoint of the update rule program and hence they are not considered in the proof of the equivalence between the fixpoints of the update rule program and the stable models for the corresponding normal logic program.

introducing the r-cliques or the a-cliques. Lemmas 4.3, 4.4 and 4.5 place restrictions on the atoms in the maximal sets, so that they may only be distinguished in certain rules in  $UR_{i+1}$  and they are not retractable in  $UR_j$ ,  $j > i+1$ . Following these restrictions, the only rules of the stratified logic program  $\overline{UR}^-(0,i+1)$  that will be modified in the normal logic program  $\overline{UR}(0,i+1)$ , through the introduction of the a-cliques or the r-cliques in this partition, are those rules that are relevant to the atoms in the maximal sets. The **Inductive Case** will thus be presented independently for the a-cliques and the r-cliques, respectively, and for each we will present a (nested) base case (no overlaps) and an inductive case (with overlaps).

### Nested base case for a-cliques

We consider one (or more) non-overlapping a-clique, comprising a-update rules  $s_1$  through  $s_p$ , with  $p$  atoms in the maximal set  $\{P_1, \dots, P_p\}$ , in the program. The rules are as follows:

$$s_k: \neg P_1, \dots, \neg P_{k-1}, \neg P_{k+1}, \dots, \neg P_p, \text{Conj-}k \rightarrow \text{assert } P_k$$

The corresponding normal logic program  $\overline{UR}(0,1)$  is in Lemma 6.3. Since the cliques are non-overlapping, they may each be considered separately, and we base the analysis on the following four sub-cases:

#### Sub-case B.a.1:

Suppose  $EDB_{i+1}^-$  entailed one (or more)  $P_k$ . Then from Lemma 4.4, since this atom  $P_k$  is unretractable, we have  $EDB_1$  will also entail this atom.  $P_k$  will be true in each fixpoint. If we consider the corresponding logic program of Lemma 6.3, the rule(s) in the program  $\overline{UR}^-(0,i+1)$  that prove  $P_k$ , or the fact  $P_k$ , and the corresponding rule(s) in any  $\overline{UR}(0,i+1)_M$ , transformed wrt some stable model  $M$ , will not be affected by the introduction of the rule(s) derived from the a-update rule(s)  $s_k$ . Thus,  $P_k$  will be true in each of the stable models for  $\overline{UR}(0,i+1)$ .

#### Sub-case B.a.2:

Suppose  $EDB_i$  does not entail some  $P_k$  occurring in this maximal set. Further, suppose that some distinguished atom  $C_k$  in the corresponding conjunction  $\text{Conj-}k$  in  $s_k$  is not entailed by  $EDB_{i+1}^-$  and is false in  $M_{\overline{UR}^-(0,i+1)}$ . Finally, suppose this  $C_k$  does not have a positive dependency on any atom in the maximal set of any a-clique. Alternately, suppose some distinguished literal  $D_k$  that occurs negatively in  $\text{Conj-}k$  is entailed by  $EDB_{i+1}^-$  and is true in  $M_{\overline{UR}^-(0,i+1)}$ . From Lemma 4.5, these literals in  $\text{Conj-}k$  are unretractable in  $UR_1$ . Now  $s_k$  will not execute and both  $EDB_{i+1}^-$  and  $EDB_{i+1}$  will not entail this  $P_k$  and it will be false in each fixpoint. In this case, the rule(s)  $r_k$  corresponding to  $P_k$  will either be eliminated in the corresponding program  $\overline{UR}(0,i+1)_M$ , transformed wrt some stable model  $M$ , (if  $D_k$  is true), or it will not be able to prove  $P_k$  (if  $C_k$  is false). It follows that  $P_k$  will be false in any stable model for  $\overline{UR}(0,i+1)$ .

#### Sub-case B.a.3:

Suppose  $EDB_i$  does not entail some  $P_k$  in this maximal set. Further, suppose that some distinguished atom  $C_k$  in the corresponding conjunction  $\text{Conj-}k$  in  $s_k$  is not entailed by  $EDB_{i+1}^-$  and is false in  $M_{\overline{UR}^-(0,i+1)}$ . Finally, suppose that  $C_k$  has a *positive dependency* on some atom which occurs in the maximal set of this a-clique, and which is not entailed by  $EDB_{i+1}^-$ . For simplicity, we assume that all other literals in  $\text{Conj-}k$  may be ignored (although this is not necessary for the proof). Suppose this atom  $C_k$ , which is distinguished in  $\text{Conj-}k$ , has a positive dependency on  $P_k$ , i.e., the atom in the head of the same rule. Then,  $s_k$  will not execute.  $EDB_1$  will not entail  $C_k$  or  $P_k$ . If we examine the corresponding stable models for  $\overline{UR}(0,i+1)$ ,  $C_k$  cannot be true in any (minimal) stable model. Any model in which  $P_k$  is true will be unstable, since  $C_k$  is false. Thus, both  $P_k$  and  $C_k$  must be false in every stable model. Alternately, suppose  $C_k$  has a positive dependency on some  $P_t$  that occurs in the maximal set,  $t \neq k$ . In this case, too,  $s_k$  cannot execute since some other rule(s)  $s_t$  in the a-clique must execute first so that  $EDB_1$  entails  $P_t$ . But from Lemma 4.6, if some  $s_t$  executes, then  $s_k$  cannot execute. Thus,  $EDB_{i+1}$  will not entail  $P_k$  but may entail  $C_k$  and  $P_t$ ,  $t \neq k$ . If we examine the corresponding stable models for  $\overline{UR}(0,i+1)$ , any model in which  $P_k$  is true must also include  $C_k$  and  $P_t$ ,  $t \neq k$ , to be stable. But such

a model in which both  $P_k$  and some  $P_t$ ,  $t \neq k$ , are true is not a minimal model for the corresponding program (from Lemma 6.3). Thus,  $P_k$  will not be true in any stable model.

#### Sub-case B.a.4:

One sub-case of interest is where  $EDB_{i+1}^-$  does not entail each of the  $P_k$ . Next, suppose that for at least one conjunction  $Conj-k$ , all distinguished atoms are entailed by  $EDB_{i+1}^-$  and are true in  $M_{\overline{UR(0,i+1)}}$  and all distinguished literals that occur negatively are not entailed and are false. It follows that the corresponding a-update rule  $s_k$  can execute and  $EDB_{i+1}$  will entail this  $P_k$ . Now since  $P_k$  is distinguished and occurs negatively in all other a-update rules participating in this a-clique, it follows that none of the other a-update rules may execute. The interesting case is where we can ignore all the conjunctions  $Conj-k$  (as was done in the proof of Lemma 6.3). If all of the a-update rules may execute, then there are  $p$  possible fixpoints for  $UR_{i+1}$ , corresponding to this a-clique. In each of the fixpoints exactly one of the  $P_k$  will be true and all other atoms in the maximal set will be false. If we consider the results of Lemma 6.3, we see that these fixpoints correspond exactly to the  $p$  possible stable models for the logic program  $\overline{UR(0,i+1)}$ .

#### Sub-case B.a.5

A final situation is to suppose that  $C_k$ , some distinguished atom in  $Conj-k$ , has a positive dependency on some atom that occurs in the maximal set of some other non-overlapping a-clique. This situation differs from B.a.4 in that  $EDB_i$  may not entail this  $C_k$  but it may be entailed in  $EDB_{i+1}$ , after introducing the non-overlapping a-clique. If  $EDB_{i+1}$  indeed entails  $C_k$ , then this sub-case will be similar to sub-case B.a.4, where the rule  $s_k$  may execute. If  $EDB_{i+1}$  does not entail  $C_k$ , then it is similar to sub-case B.a.2, where the rule  $s_k$  cannot execute.

### Nested inductive case for a-cliques

Suppose the fixpoints for programs with  $n$  a-cliques that overlap in  $UR_{i+1}$  entail all possible stable models for the corresponding Gelfond-Lifschitz transformed programs. Now consider introducing another a-clique which overlaps with the previously overlapping  $n$  a-cliques, i.e., there are  $n+1$  overlapping a-cliques. Each of the overlaps of this clique with one of the  $n$  other cliques may be represented as a set of a-update rules  $s_1$  through  $s_{p+c}$ , relevant to  $p+c$  atoms  $\{P_1, \dots, P_p, C_1, \dots, C_c\}$  and a set of a-update rules  $u_1$  through  $u_{q+c}$ , relevant to  $q+c$  atoms  $\{Q_1, \dots, Q_q, C_1, \dots, C_c\}$ .  $\{C_1, \dots, C_c\}$  represent the overlapping atoms common to the maximal sets, for each of the overlaps with the  $n$  a-cliques, respectively. The rules are as follows, where  $s_{k1}$  is representative of rules  $s_1$  through  $s_p$ , and  $s_{p+k3}$  is representative of rules  $s_{p+1}$  through  $s_{p+c}$ , etc. :<sup>1</sup>

$$s_{k1}: \neg P_1, \dots, \neg P_{k1-1}, \neg P_{k1+1}, \dots, \neg P_p, \neg C_1, \dots, \neg C_c, Conj-s,k1 \rightarrow \mathbf{assert} P_{k1}$$

$$s_{p+k3}: \neg P_1, \dots, \neg P_p, \neg C_1, \dots, \neg C_{k3-1}, \neg C_{k3+1}, \dots, \neg C_c, \neg Q_1, \dots, \neg Q_q, Conj-s,p+k3 \rightarrow \mathbf{assert} C_{k3}$$

$$u_{k2}: \neg Q_1, \dots, \neg Q_{k2-1}, \neg Q_{k2+1}, \dots, \neg Q_q, \neg C_1, \dots, \neg C_c, Conj-u,k2 \rightarrow \mathbf{assert} Q_{k2}$$

$$u_{q+k3}: \neg Q_1, \dots, \neg Q_q, \neg C_1, \dots, \neg C_{k3-1}, \neg C_{k3+1}, \dots, \neg C_c, \neg P_1, \dots, \neg P_p, Conj-u,q+k3 \rightarrow \mathbf{assert} C_{k3}$$

The corresponding normal logic program  $\overline{UR(0,i+1)}$ , for each of these  $n$  overlaps, is in Lemma 6.4. We consider several sub-cases as follows:

#### Sub-case I.a.1

In this sub-case, each of the a-update rules  $s_1$  through  $s_{p+c}$  and  $u_1$  through  $u_{q+c}$  do not execute. The conditions for these rules to not execute are already detailed in sub-cases B.a.1 through B.a.3, and B.a.5, of the base case. There is a slight variation that must be considered in sub-case B.a.5, since now there are overlaps among the

<sup>1</sup> Note that the a-update rules that are relevant to the overlapping atoms in the maximal sets are repeated in this program and should not be considered to be different rules.

a-cliques. For this case, we may suppose that  $C_{k1}$ , some distinguished atom in  $Conj-s,k1$ , has a positive dependency on either some atom  $Q_{k2}$  (which is not an atom in the overlap) or on some atom which occurs in the maximal set of some other a-clique with which there is no overlap at all. In all of these sub-cases, the overlap with each of the  $n$  a-cliques is itself not significant, and the analysis is similar to the base case for a-cliques. We omit this discussion here.

There are several sub-cases of interest where the a-update rules may execute, and they are as follows:

**Sub-case I.a.2:**

Suppose  $EDB_{-i+1}^-$  does not entail each of the  $P_{k1}$  and  $C_{k3}$  occurring in the maximal set for rules  $s_1$  through  $s_{p+c}$ . Further, suppose that we can ignore each of the conjunctions in the a-update rules relevant to each  $P_{k1}$  and each  $C_{k3}$ , i.e., all distinguished atoms are entailed by  $EDB_{-i+1}^-$  and are true in  $M_{\overline{UR(0,i+1)}}$ , and all distinguished literals occurring negatively are not entailed and are false. Thus, the rules  $s_1$  through  $s_{p+c}$  may execute. However, the a-update rules  $u_1$  through  $u_{q+c}$  may not execute (similar to the situation described in one of the previous sub-cases of the base case (B.a.1 through B.a.3 and B.a.5)).

**Sub-case I.a.3:**

Suppose the rules  $s_1$  through  $s_{p+c}$  may not execute (similar to one of the sub-cases B-a.1 through B.a.3 and B.a.5) but the rules  $u_1$  through  $u_{q+c}$  may execute.

Each of these two sub-cases will now reduce to  $n$  overlapping a-cliques and the result follows from the statement of the inductive case for  $n$  overlapping a-cliques.

**Sub-case I.a.4:**

Suppose  $EDB_{-i+1}^-$  does not entail each of the  $P_{k1}$ ,  $Q_{k2}$  and  $C_{k3}$  occurring in both maximal sets. Further, suppose that we can ignore each of the conjunctions in the a-update rules relevant to each  $P_{k1}$ ,  $Q_{k2}$  and each  $C_{k3}$ . We made this assumption when proving the results of Lemma 6.4. Now any of the a-update rules in the overlapping a-cliques may execute. Suppose an a-update rule  $s_{k1}$  were to execute and entail some  $P_{k1}$ . From lemma 4.6, no other a-update rule  $s_t$ ,  $t \neq k1$ , could execute and no other  $P_t$ ,  $t \neq k1$ , may be entailed. Similarly, no  $C_t$  may be entailed. However, exactly one a-update rule  $u_{k2}$  can execute and entail exactly one  $Q_{k2}$ . Suppose on the other hand that some a-update rule relevant to some  $C_{k3}$  were to execute and entail some  $C_{k3}$ . Then, from Lemma 4.6, no other  $C_t$ ,  $t \neq k3$ , may be entailed nor any other  $Q_t$  or  $P_t$ . It follows that there are exactly  $(p * q + c)$  possible fixpoints, exactly corresponding to the  $(p * q + c)$  stable models of the program  $UR(0,i+1)$ , as seen in Lemma 6.4.

**Nested base case for r-cliques**

We consider one (or more) non-overlapping r-clique with  $p$  retractable atoms in the maximal set  $\{P_1, \dots, P_p\}$ . The  $p$  r-update rules  $s_1, \dots, s_p$  participating in the r-clique are as follows:

$$s_k: P_1, \dots, P_k, \dots, P_p, Conj-s,k \rightarrow \text{retract } P_k$$

The a-update rules  $p_k$  which are relevant to the retractable atoms in the maximal set of atoms associated with the r-clique and are modified by these r-update rules are as follows:

$$p_k: Conj-p,k \rightarrow \text{assert } P_k$$

From Lemma 4.2, we know that these a-update rules may not participate in an a-clique. The corresponding normal logic program  $\overline{UR(0,i+1)}$  is in Lemma 6.1. Since the cliques are non-overlapping, each can be analyzed separately, based on the following sub-cases:

**Sub-case B.r.1:**

Suppose that some distinguished atom  $C_k$  in  $Conj-s,k$  is not entailed by  $EDB_{-i+1}^-$  and is false in  $M_{\overline{UR(0,i+1)}}$ . Further suppose that this atom  $C_k$  does not occur in the maximal set of any a-clique in  $UR_{i+1}$ , nor does it have

a positive dependency on any other atom that occurs in a maximal set of an a-clique. Alternately, suppose some distinguished literal  $D_k$  that occurs negatively is entailed by  $EDB_{i+1}^-$  and is true in  $M_{\overline{UR(0,i+1)}}$ . From Lemma 4.5, these literals are unretractable in  $UR_{i+1}$ . In this case, the corresponding r-update rule(s)  $s_k$  will not be able to execute. If  $EDB_{i+1}^-$  entailed the corresponding atom  $P_k$ , then,  $EDB_{i+1}$  will also entail this atom.

If we consider the corresponding logic program of Lemma 6.1, and the program  $\overline{UR(0,i+1)}_M$  transformed wrt to any stable model  $M$ , the rule(s)  $r_{1,k}$ , derived from  $s_k$ , will either be eliminated (if  $D_k$  is true), or it will not be able to prove  $P_k^*$  (if  $C_k$  is false).  $P_k^*$  will be false in any stable model. Consider the rule(s)  $r_{2,k}$  in  $UR^-(0,i+1)$ , relevant to  $P_k$  and which is derived from  $p_k$  transformed by  $s_k$ . It will be unchanged in any transformed program  $\overline{UR(0,i+1)}_M$ , since the literal  $P_k^*$  occurring negatively in this rule will be eliminated. We do not further consider this case since the fixpoint  $EDB_{i+1}$  remains unchanged from  $EDB_{i+1}^-$  wrt this atom  $P_k$ , and the corresponding rules relevant to  $P_k$ , in any  $\overline{UR(0,i+1)}_M$ , remain unchanged from  $UR^-(0,i+1)$ .

#### Sub-case B.r.2:

Similarly, suppose  $EDB_i$  does not entail some  $P_k$  occurring in this maximal set. Further, suppose that some distinguished atom  $C_k$  in the corresponding conjunction  $Conj-p,k$  in  $p_k$  is not entailed by  $EDB_{i+1}^-$  and is false in  $M_{\overline{UR(0,i+1)}}$ . Again, we assume the same restrictions for  $C_k$  as in B.r.1. Alternately, suppose some distinguished literal  $D_k$  that occurs negatively is entailed and is true. Now,  $p_k$  will not execute, and both  $EDB_{i+1}^-$  and  $EDB_{i+1}$  will not entail this  $P_k$ . The corresponding rule(s)  $r_{2,k}$  will not be able to prove  $P_k$  in  $UR^-(0,i+1)$ . Also, in any  $\overline{UR(0,i+1)}_M$ , transformed wrt any stable model  $M$ , rule(s)  $r_{2,k}$  will either not be able to prove  $P_k$  or will be eliminated. Thus, this case is similar to the previous case B.r.1.

#### Sub-case B.r.3:

One case of interest is where we can first ignore all the conjunctions  $Conj-p,k$ . This was one of the assumptions made to prove the results of Lemma 6.1. First, suppose all distinguished atoms in all such conjunctions  $Conj-p,k$  are entailed by  $EDB_{i+1}^-$  and are true in  $M_{\overline{UR(0,i+1)}}$  and all distinguished literals that occur negatively in these conjunctions are not entailed and are false. Now  $EDB_{i+1}^-$  will entail all  $P_k$  and they will all be true in  $M_{\overline{UR(0,i+1)}}$ . Next, suppose that for at least one conjunction  $Conj-s,k$ , all distinguished atoms are entailed by  $EDB_{i+1}^-$  and are true in  $M_{\overline{UR(0,i+1)}}$  and all distinguished literals occurring negatively are not entailed and are false. It follows that the corresponding r-update rule  $s_k$  can execute and  $EDB_{i+1}$  will not entail  $P_k$ . From lemma 4.6, none of these other r-update rules  $s_t$ ,  $t \neq k$ , can execute. Thus,  $EDB_{i+1}$  will not entail some  $P_k$  and will entail all other  $P_t$ ,  $t \neq k$ .

In the most general case, suppose that for all the r-update rules participating in the r-clique, we can ignore all the conjunctions,  $Conj-s,k$  (as was done. in Lemma 6.1). It follows that any of the r-update rules participating in this r-clique may execute and from Lemma 4.6, only one r-update rule in the r-clique will execute. There will be  $p$  possible fixpoints represented by some  $EDB_{i+1}$ . Each fixpoint will not entail exactly one  $P_k$  and will entail all other  $P_t$ ,  $t \neq k$ .

From Lemma 6.1, when we correspondingly ignore all the literals in the conjunctions  $Conj-s,k$  and  $Conj-p,k$ , we have that there are exactly  $p$  stable models for the corresponding normal logic program  $\overline{UR(0,i+1)}$ . In each of these stable models, exactly one retractable atom of the maximal set of atoms will be false and all other retractable atoms will be true, where we ignore the special literals  $P^*$  introduced into the normal logic program. Thus, each possible fixpoint exactly entails one possible stable model and vice versa, ignoring the special literals introduced into the normal logic program.

#### Sub-case B.r.4:

Suppose that some distinguished atom  $X_k$  in  $Conj-s,k$  in some rule  $s_k$  (or in  $Conj-p,k$  in some rule  $p_k$ ) is not entailed by  $EDB_{i+1}^-$  and is false in  $M_{\overline{UR(0,i+1)}}$ . Further suppose this atom  $X_k$  either occurs in the maximal set



of some a-clique in  $UR_{i+1}$ , or it has a positive dependency on some other atom that occurs in a maximal set of an a-clique. Now, after executing the rules in the a-clique,  $EDB_{i+1}$  may entail  $X_k$ . This sub-case is different from B.r.1 and B.r.2, in that  $EDB_{i+1}^-$  will not entail  $X_k$ , but  $X_k$  may be entailed by the fixpoint  $EDB_{i+1}$ . The rules  $s_k$  (or  $p_k$ ) may or may not execute, depending on the sequence of execution of the rules, when the rules in the a-cliques and the r-cliques are introduced. For example, if a rule in some a-clique entailed  $X_k$ , then  $p_k$  alone or  $s_k$  and  $p_k$  may execute. However, if some other rule  $s_t$   $t \neq k$ , in the r-clique, executes first, then  $s_k$  will not execute. If some rule in the r-clique does not execute, then this sub-case will be similar to B.r.1 or B.r.2, and if some rule does execute, then this sub-case will reduce to some form of the previous sub-case B.r.3.

### Nested inductive case for a-cliques

Suppose the fixpoints for a programs with n r-cliques that overlap in  $UR_{i+1}$  entail all possible stable models for the corresponding Gelfond-Lifschitz transformed programs. Now consider introducing another r-clique which overlaps with the previously overlapping n r-cliques, i.e., there are n+1 overlapping r-cliques. Each of the overlaps of this r-clique with one of the n other cliques may be represented as a set of r-update rules  $u_1$  through  $u_{p+c}$ , relevant to p+c atoms  $\{P_1, \dots, P_p, C_1, \dots, C_c\}$  and a set of r-update rules  $s_1$  through  $s_{q+c}$ , relevant to q+c atoms  $\{Q_1, \dots, Q_q, C_1, \dots, C_c\}$ .  $\{C_1, \dots, C_c\}$  represent the overlapping atoms common to the maximal sets, for each of the overlaps with the other n r-cliques. These overlapping r-cliques are as follows: <sup>1</sup>

$$\begin{aligned} u_{k1}: & P_1, \dots, P_{k1}, \dots, P_p, C_1, \dots, C_c, \text{Conj-}p,k1 \rightarrow \mathbf{retract} P_{k1} \\ u_{k3}: & P_1, \dots, P_p, C_1, \dots, C_{k3}, \dots, C_c, Q_1, \dots, Q_q, \text{Conj-}u,k3 \rightarrow \mathbf{retract} C_{k3} \\ s_{k2}: & Q_1, \dots, Q_{k2}, \dots, Q_q, C_1, \dots, C_c, \text{Conj-}s,k2 \rightarrow \mathbf{retract} Q_{k2} \\ s_{k3}: & Q_1, \dots, Q_q, C_1, \dots, C_{k3}, \dots, C_c, P_1, \dots, P_p, \text{Conj-}s,k3 \rightarrow \mathbf{retract} C_{k3} \end{aligned}$$

The a-update rules which are relevant to these retractable atoms and are modified by the r-update rules are as follows:

$$\begin{aligned} p_{k1}: & \text{Conj-}p,k1 \rightarrow \mathbf{assert} P_{k1} \\ p_{k2}: & \text{Conj-}p,k2 \rightarrow \mathbf{assert} Q_{k2} \\ p_{k3}: & \text{Conj-}p,k3 \rightarrow \mathbf{assert} C_{k3} \end{aligned}$$

From Lemma 4.2, these a-update rules do not participate in an a-clique. The corresponding normal logic program is as listed in Lemma 6.2. As in the analysis of the previous **nested base case** for r-cliques, there are several situations that are not of much interest. For example, when all of the a-update rules and r-update rules cannot execute (detailed in B.r.1 and B.r.2), the overlap with each of the n a-cliques is itself not significant. Similarly, if we can ignore all the a-update rules and r-update rules relevant to some overlapping maximal set, then this reduces to the case of n overlapping r-cliques and the result follows from the statement of the nested inductive case for n overlapping cliques. We do not list these sub-cases here for readability.

#### Sub-case I.r.1:

One interesting case is to suppose  $EDB_{i+1}^-$  will entail all  $P_k$  and all  $C_m$  as well as all  $Q_i$  and they will all be true in  $M_{UR-(0,i+1)}$ . Once more we suppose that we can ignore some (or all) of the conjunctions  $\text{Conj-}u,k$ ,  $\text{Conj-}u,m$ ,  $\text{Conj-}s,t$  or  $\text{Conj-}s,m$ . <sup>1</sup>

Now either exactly one  $r_{k1}$  and exactly one  $s_{k2}$  will execute and  $EDB_{i+1}$  will not entail exactly one  $P_{k1}$  and exactly one  $Q_{k2}$ . An alternative is that exactly one r-update rule  $r_{k3}$  (or  $s_{k3}$ ) will execute and  $EDB_{i+1}$  will not entail exactly one  $C_{k3}$ . In the most general case, suppose that any of the r-update rules participating in the

<sup>1</sup> Note that the r-update rules that are common to the intersecting r-cliques are repeated.

mutually overlapping  $r$ -cliques may be chosen for execution. Then, since there are  $(p+c)$  and  $(q+c)$  atoms in the maximal sets, respectively, with  $c$  mutually overlapping literals, there will be  $(c + p * q)$  possible fixpoints, in this most general case.

We note that we ignored the various conjunctions in the rules of the corresponding normal logic program in the proof of Lemma 6.2. We further note from Lemma 6.2, that the  $(p * q + c)$  possible fixpoints that we have enumerated for the most general case exactly correspond to the stable models for the normal logic program, ignoring all the special literals introduced into the normal logic program. We refer the reader to Lemma 6.2 for the details.

### Sub-case I.r.2:

The final sub-case of interest is where some distinguished atom  $X_k$  in one of the  $a$ -update rules (that do not participate in any  $a$ -clique) or any of the  $r$ -update rules of any  $r$ -clique, is not entailed by  $EDB_{i+1}^-$  and is false in  $M_{UR(0,i+1)}$ . However, suppose this atom  $X_k$  either occurs in the maximal set of some  $a$ -clique in  $UR_{i+1}$ , or it has a positive dependency on some other atom that occurs in a maximal set of an  $a$ -clique. The execution of the rules in this  $a$ -clique may entail  $X_k$ . This sub-case is similar to B.r.4 in that  $EDB_{i+1}^-$  will not entail  $X_k$ , but  $X_k$  may be entailed in the fixpoint  $EDB_{i+1}$ . As before the  $r$ -update rules in the  $r$ -cliques and the  $a$ -update rules (that do not participate in any  $a$ -cliques) may or may not execute, depending on the execution sequence of the rules in the  $a$ -cliques and the  $r$ -cliques. This case may be similar to I.r.1 if  $X_k$  is entailed or it will reduce to  $n$  overlapping  $r$ -cliques if  $X_k$  is not entailed. We omit the detailed discussion.  $\square$

## 7. Comparisons with Related Research

We now compare our research with related research in providing a semantics for update rules as well as research on implementing rules in a DBMS. There has been considerable research to define a correct semantics for update rule programs. In [FKUV86], a general framework for dealing with updates is presented. The problem they consider is updating a theory by inserting and/or deleting sentences. A theory  $T$  is defined as a consistent set of sentences, in particular, a set of first-order, well-formed closed formula. The set of logical consequences of the theory is denoted  $T^*$ . Let  $\sigma$  be a sentence. Then, the theory  $S$  accomplishes the update corresponding to the insertion of  $\sigma$  into  $T$  if  $\sigma$  is an element of  $S$ . Similarly,  $S$  accomplishes the update corresponding to the deletion of  $\sigma$  from  $T$  if  $\sigma \notin S^*$ . Notice that the inserted sentence is explicit in  $S$ . Now suppose that  $T \cup \sigma$  is inconsistent but  $\sigma$  is to be inserted, or that  $\sigma$  is to be deleted while  $\sigma \in T^*$ . In both of these cases, a *minimal* new theory  $S$  must be defined. Suppose that  $T_1$  and  $T_2$  are both theories that accomplish the update. Then, the authors define what it means for  $T_1$  to accomplish the update with a *smaller change* than  $T_2$ . Now,  $S$  accomplishes an update  $u$  of  $T$  *minimally* if there is no theory  $S'$  which accomplishes update  $u$  with a *smaller change* than  $S$ .

The research in [FKUV86] provides a conceptual framework for defining a semantics for updating sentences in a theory. The problem we address is a simpler problem since our database corresponds to a set of facts. The UR program comprises an initial database of facts  $EDB_{init}$  and a set of update rules. Intuitively, the  $a$ -update rules that assert new facts are similar to deductive rules since each rule asserts new information based on some condition that is satisfied by the database. The  $r$ -update rules that retract facts intuitively resemble integrity constraints. They check the database for conditions that must not hold. If the condition does hold, then they retract some information that exists in the database so the condition no longer holds. This is very similar to the process of maintaining integrity constraints in a DBMS environment. Our objective then is to obtain a final database for the initial facts and the new facts that are asserted such that this final database is minimal, and is consistent with the conditions specified by the  $r$ -update rules, representing the integrity constraints.

Our approach is different from the framework suggested in [FKUV86] since we treat the r-update rules as integrity constraints. The final database must satisfy these integrity constraints. As a result, information which is explicitly specified by the user in the initial database  $EDB_{init}$  may be later retracted by some update rule, in order to maintain an integrity constraint. The same is true for information which is explicitly deleted by the user since it may be asserted by another rule. In contrast, in [FKUV86], information which is explicitly updated or retracted by the user is not changed. Another feature of our fixpoint semantics is that when a tuple is retracted by a rule, then subsequent assertions (of the tuple) have no effect and the database will not entail this tuple.

To obtain a declarative semantics, we associate a normal logic program,  $\overline{UR}$ , with a UR program comprising of the initial database  $EDB_{init}$  and the update rules. The stable model semantics for normal logic programs [GeLi88] associates a set of minimal models which are called stable models with a normal logic program  $\overline{UR}$ . A stable model for  $\overline{UR}$  must be equivalent to the database obtained through the execution of the update rules of UR. The stable model is consistent with the conditions specified by the integrity constraints. This approach to providing a declarative semantics based on the stable model semantics of a corresponding normal logic program is also different from other research in this area.

The most extensive research in update rules has been described in [AbSV90, AbVi90, AbVi91]. They extend Datalog programs together with a procedural component as a means of overcoming the limited expressive power of purely declarative semantics, for example, the inability to capture updates in pure Datalog programs. The focus is on the expressive power of the languages. Theoretical results on computational complexity are discussed. They also investigate the connections between Datalog extensions with fixpoint semantics, explicitly procedural languages and fixpoint extensions of first order logics. The Datalog extensions that are considered include negative literals in the body of the rules as well as in the heads. They define invented values which correspond to free variables occurring in the heads of rules. Negative literals in the heads of rules are interpreted as deletions. Both deterministic and non-deterministic semantics for Datalog programs extended with negation are discussed. In a deterministic execution, all the rules whose bodies are satisfied in the database will be chosen simultaneously, and their updates applied to the database, so that the order of executing rules is not significant. The issue of applying the updates of more than one rule simultaneously has also been addressed by other researchers in the context of set-oriented execution, as will be discussed.

In our research, we are interested in a much smaller class of update rules whose fixpoint semantics can be correctly implemented in a relational DBMS with extensions. Another difference is that update rules which delete facts are interpreted by us as integrity constraints. This is intuitively closer to the meaning commonly associated with rules that delete facts, from the perspective of a designer in a DBMS environment. This contrasts with their solution of equating deletions with negative conclusions in the heads of rules. Our fixpoint semantics is non-deterministic in that the fixpoint operator selects a single rule from a partition, in each step. Since the order of rule selection is significant, it could produce more than one final database.

The research most similar to our approach is described in [MaSi88, SiMa88]. They consider production rule languages that are extensions of logical query languages such as Datalog with updates in the head of the rules. They provide an operational and declarative semantics for stratified Datalog programs that have a sequence of updates in the heads of the rules. Their declarative semantics is similar in spirit to modal logics. In addition to an implicit ordering of the rules due to stratification, explicit ordering is also considered. In our research, we have extended the concept of stratification for update rules to permit us to represent non-deterministic programs, but we use no explicit control. Their execution model is based on a Predicate Compilation Network (PCN) and is derived from Petri-Net models. One drawback is that in the PCN model for rule execution, there may not be a stable state associated with the network representing the rule programs; this implies that rule execution may not terminate. They use explicit control during rule execution to deal with this problem. Our execution semantics is based upon a monotonic fixpoint operator. Asserts and retracts are explicitly stored and the fixpoint operator guarantees termination of rule

execution. We do not require any explicit control information for the execution to terminate.

We note that although we have not considered programs with multiple updates in the head of a rule, this is a straightforward extension. Another difference from the research in [MaSi88] is that we interpret update rules which delete information as integrity constraints; we feel that this is more intuitive with the use of these rules from the viewpoint of DBMS designers. We are also interested in identifying a class of programs that can be easily implemented in a DBMS. Executing a rule must be very similar to query execution so that all the techniques developed for the efficient execution of queries is applicable to rule execution as well.

There has been some research in update semantics in conjunction with logic programming research [MaWa88, NaKr88]. In the language proposed in [MaWa88], updates can occur in the body of the rules, as well as in the heads. In the language proposed in [NaKr88], the updates occur only in the body of the rules. The declarative semantics for both proposals are based on modal logics. The focus of their research is in providing a formal semantics; implementations are based on a proof theory, which is not as applicable in a DBMS environment where the rule execution is usually implemented to be a forward chaining evaluation.

Finally, we discuss related research in [CeWi90, Wido91, WiFi90], which describes implementing production rules in a DBMS environment. A very powerful set-oriented production rule language is defined and an execution semantics which is also set-oriented is provided in [WiFi90]. Since the language is set-oriented, there is a related issue of whether a tuple-oriented and a set-oriented implementation will provide the same or different answers; this is an important issue if a precise semantics for the program is to be defined. We note that although the set-oriented execution described in [WiFi90] will apply the changes of multiple rules simultaneously, it does not necessarily produce a deterministic answer, and the order of execution of each rule is significant. Their semantics is therefore different from the deterministic semantics of [AbSV90]. If we consider our rule language, the syntactic restrictions we place on the UR programs guarantees that a set-oriented execution is a special case of the tuple-oriented execution. We see this as an advantage since the set-oriented execution is similar to query processing in a DBMS and can be efficient.

The focus of the research in [Wido91, WiFi90] is on the computational procedures required to determine when to trigger rule execution, and determining when no more rules can be executed. They have identified the concept of a *transition table* to represent all the tuples which are affected by executing a transaction; the transition table is critical in determining when the rules are to be evaluated. These are important implementation issues. The transition table also allows representation of a change in state of the database after rule execution, and thus represents a very different semantics from our research.

In [CeWi90], a very powerful constraint language is defined and a procedure for deriving constraint maintaining production rules is described. Their research focus is on the nature of constraints and the process of maintaining constraints. We do not explore these issues in our paper. They guarantee that if the correct actions are specified, corresponding to each violation of a constraint, then the rules are guaranteed to take the corrective actions, as specified. In our research, too, we treat the update rules which retract information as rules that maintain an integrity constraint. However, in our research, we use the declarative meaning given to the program comprising the initial database and the update rules to determine which rules must execute and to determine the final database. This is an advantage since we provide a declarative meaning for the fixpoint or forward-chaining semantics. An interesting issue for future research would be to consider using our update rules and the fixpoint semantics to implement the constraint language defined in [CeWi90].

## 8. Implementing Update Rules in a Relational DBMS

We briefly outline an implementation of the fixpoint semantics for update rule programs in an extended DBMS environment. An important requirement is that the implementation should be straightforward, and require minimal extensions to the relations or to the query processing strategy in a DBMS. We describe an incremental

evaluation scheme where each rule is represented by a query which conditionally updates the relations in the database.

### 8.1. Obtaining Queries from the Rules

We assume there is a relation corresponding to each predicate. The number and type of the attributes of the relations correspond to the number and type of the terms in the predicate. The *database* relations correspond to predicates which do not occur in the head of any update rules. The *updatable* relations correspond to predicates which occur in the head of the update rules. These predicates may also occur in the body of the rules. To represent tuples that are asserted and retracted, while guaranteeing that rule execution will terminate, each updatable relation is extended with two additional attributes. These two attributes are named the **a-flag** and **r-flag**, respectively. The update rules are represented by a set of queries that execute against both the database and the updatable relations.

When a tuple is inserted into an updatable relation through an explicit insert operation or when it is asserted through the execution of an a-update rule, then the corresponding tuple is placed in the updatable relation. It has its **a-flag** value set to **1**, and its **r-flag** value set to **0**, to indicate it has been asserted. When this tuple is retracted from an updatable relation, by executing an r-update rule, the value of the **r-flag** is set to **1**. When a tuple is deleted from an updatable relation explicitly, then it will be actually deleted from this relation. Inserts and deletes from the database relations are as usual, i.e., these relations are not extended with the flags.

Literals that correspond to database relations are interpreted in the usual manner against the tuples in the database relations. Informally, a *unifying* tuple in a database relation *satisfies* a positive literal when the attribute values of the unifying tuple satisfy the selection predicate that is specified in the vector of terms (constants and/or variables), corresponding to that literal. The exact relationship between the vector of terms in the literal and the attributes of the unifying tuple is straightforward. A literal occurring negatively in the body of a rule, which corresponds to a database relation, is satisfied when there are no unifying tuples in the corresponding relation that satisfy the selection predicate.

However, for literals corresponding to updatable relations, there are additional requirements that must be met for the literals to be satisfied. They are as follows:

- (1) A literal occurring positively in the body of a rule will be satisfied by the existence of a unifying tuple of an updatable relation which satisfies the selection predicate and has its **a-flag** value equal to 1 and its **r-flag** value equal to 0. This indicates that the unifying tuple has either been explicitly inserted into the relation or it has been asserted by an a-update rule, but it has not been retracted by an r-update rule.
- (2) A literal occurring negatively in the body of a rule will be satisfied if *all* unifying tuples of an updatable relation that satisfy the selection predicate have both the **a-flag** value and the **r-flag** value set equal to 1.
- (3) A literal occurring negatively in the body of a rule will also be satisfied when there are no unifying tuples in the corresponding updatable relations, or all the unifying tuples do not satisfy the selection predicate.
- (4) A literal occurring negatively in the body of a rule will *not* be satisfied if there exists *any* unifying tuple which matches the selection predicate and has its **a-flag** set equal to 1 but its **r-flag** set equal to 0. This tuple has been explicitly inserted or asserted by an a-update rule but it has not been retracted by an r-update rule or explicitly deleted.

The details of obtaining the corresponding queries from the update rules, based on the above criteria, is straightforward. When these queries are executed, unifying tuples are retrieved from the relations corresponding to the literals in the body of the rule, so that it can be determined if these unifying tuples satisfy the selection predicate in the literals in the body of the rules. For those rules where the literals in the body are satisfied, the queries update the updatable relation corresponding to the literal in the head of the rule.

## 8.2. Maintaining an Update Rule System in a DBMS

The first step is to create a database and populate the relations to reflect the state of the initial database. This is the *Initialize* step. Let  $EDB_{init}$  be a set of tuples representing the initial database. This step is as follows:

### Initialize

For each database relation P

For each tuple  $P(\bar{t})$  in  $EDB_{init}$

```
{ INSERT INTO P
  (P.attri := the corresponding constant in P( $\bar{t}$ )) }
```

For each updatable relation P

For each tuple  $P(\bar{t})$  in  $EDB_{init}$

```
{ INSERT INTO P
  (P.a-flag := 1) AND (P.r-flag := 0) AND
  (P.attri := the corresponding constant term in P( $\bar{t}$ )) }
```

Once a database has been created, there are three phases in the processing required to implement the semantics of the update rule system. We expect that the system will cycle through these phases each time the database relations are explicitly updated by a user, in order to maintain the updatable relations.

In the first phase, *ExplicitUpdate*, the contents of the relations must be updated to reflect the changes to both database and updatable relations, as specified explicitly by the user. Suppose  $EDB_{ins}$  are the tuples that are inserted and  $EDB_{del}$  are the tuples that are deleted. Clearly, as the database is updated over time, only a small portion of  $EDB_{init}$  may actually change in each cycle. In the *ExplicitUpdate* phase, the processing is as follows:

### ExplicitUpdate

For each database relation P

For each tuple  $P(\bar{t})$  in  $EDB_{ins}$

```
{ INSERT INTO P
  (P.attri := the corresponding constant in P( $\bar{t}$ )) }
```

For each tuple  $P(\bar{t})$  in P which is in  $EDB_{del}$

```
{ DELETE FROM P
  WHERE {for each attribute in P}
  (P.attri = the corresponding constant in P( $\bar{t}$ )) }
```

For each updatable relation P

For each tuple  $P(\bar{t})$  in  $EDB_{ins}$

```
{ INSERT INTO P
  (P.a-flag := 1) AND (P.r-flag := 0) AND
  (P.attri := the corresponding constant in P( $\bar{t}$ )) }
```

For each tuple  $P(\bar{t})$  in P which is in  $EDB_{del}$

```
{ DELETE FROM P
  WHERE
  (P.a-flag = 1) AND (P.r-flag = 0) AND
  {for each attribute in P}
  (P.attri = the corresponding constant in P( $\bar{t}$ )) }
```

The second phase is the *Evaluation*. Let  $Q-UR_i$  be the set of database queries derived from the set of a-update rules and/or r-update rules in each partition  $UR_i$  of the update rule program UR. The database queries corresponding to each partition of  $UR_i$  are executed, in turn, starting from  $UR_1$ , until the queries in each partition,  $Q-UR_i$  can no longer update the updatable relations, i.e., a fixpoint is reached.



## Evaluation

For each partition  $UR_i$  in increasing order of  $i$

Non-deterministically select and execute a query from  $Q-UR_i$   
and update the updatable relations until a fixpoint is reached

The final phase is the *Completion*. All the tuples which have been retracted by the execution of an r-update rule are deleted from the updatable relations. To explain, it is only during the processing of the *Evaluation* phase that we need to explicitly maintain information on all the tuples that were retracted through the execution of r-update rules so that the execution will terminate. Once the *Evaluation* completes, we no longer require this information about the retracted tuples, and these tuples can be deleted. This will reduce the size of the updatable relations. It will also allow a tuple which was retracted through the execution of an r-update rule in the current Evaluation phase to be re-asserted in the next ExplicitUpdate phase, without compromising the conditions for termination of rule execution. The *Completion* is as follows:

## Completion

For each updatable relation  $P$

```
{ DELETE FROM P
  WHERE
  (P.a-flag = 1) AND (P.r-flag = 1) }
```

## 9. Summary and Future Research

In this paper, we have defined a semantics for update rules, i.e., rules whose execution can update the database and can cause the further execution of rules. Our motivation is to define a class of programs which can be easily identified by a DBMS designer and which will be useful in common DBMS applications such as to capture triggering information, to maintain integrity constraints, or to make simple inferences. Our research focus is on supporting these UR programs in an extended DBMS environment. The class of UR programs is syntactically identified, based on the concept of stratification. We extend the strict definition of stratification and allow a more flexible criterion for partitioning the rules in the UR program. This relaxation allows a certain degree of non-determinism in rule execution, which can be very useful in maintaining integrity constraints in the database.

We define a semantics which is based upon a monotonic fixpoint operator  $T_{UR}$ . The monotonicity of the operator is achieved by explicitly recording the effects of inserts and deletes of the tuples in the database. The semantics guarantee the execution of the rules will terminate and will produce a minimal database.

To obtain a declarative semantics for UR programs, we associate a normal logic program  $\overline{UR}$  with each UR program. We use the stable model semantics for normal logic programs [GeLi88], which associates a set of stable models with each normal logic program. The set of stable models are equivalent to the minimal databases that are (non-deterministically) obtained in the fixpoint for the UR program. We prove that the fixpoint semantics for the update rule program is sound and complete wrt the stable model semantics for the corresponding normal logic program.

We describe implementing the semantics of the UR program in a DBMS environment. Relations that can be updated by the rules or *updatable* relations are extended with two flags. An update rule can be represented by a database query, which queries the updatable relations as well as database relations (which are not updated by the rules). We describe an algorithm to compute a fixpoint in the DBMS environment and obtain the final updated database. A prototype system that recognizes the legal update rule programs, translates the rules into database queries and executes these queries against the database relations and the updatable relations is being built.

In future research, we propose extensions to the class of UR programs, so that they may be more useful in the context of DBMS applications. This includes the support of explicit control information for rule execution. We



also wish to allow multiple updates in the head of each rule. From the viewpoint of declarative semantics, we would like to be able to extend the semantics so that the programs may be able to handle exceptions as used in AI programs, and causality [RaLo94]. We feel these extensions will be important to solve the problem of combining multiple knowledge bases.

We also propose to study the implementation issues in a DBMS more extensively. This includes the problem of incremental computation of queries, and the concurrent execution and set-oriented execution of update rules. We are also investigating magic transformations of the update rule programs to provide more efficient bottom-up computation techniques [PaRa94].

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## 11. Appendix 1: The Translation with First-Order Predicates

In the case of the first-order predicates, we first specify some syntactic restrictions on the vectors of terms, (variables and/or constants), associated with predicates in the cliques<sup>1</sup>. These restrictions ensure that the same (set of) tuples are unifiable with each occurrence of the predicates, in all of the rules participating in a clique. We then describe the translation to obtain a normal logic program.

### 11.1. Specifying the Cliques for First-Order Predicates

Suppose  $P_1, \dots, P_p$ , are the predicates that occur in some maximal set associated with either an a-clique or an r-clique. Further, suppose that the vector of terms associated with predicate  $P_k$ , in some a-update rule or r-update rule  $r_t$ , in this clique, is  $\overline{u_{k,t}}$ . The term  $\overline{u_{k,t}}$  must be unifiable with all other terms  $\overline{u_{k,s}}$ , for  $P_k$ , occurring in some other rule  $r_s$ ,  $t \neq s$ . In addition, the following conditions must hold:<sup>2</sup>

- (1) Consider any two rules  $r_{t1}$  and  $r_{t2}$  in the clique, and the terms  $\overline{u_{k,t1}}$  and  $\overline{u_{k,t2}}$ , associated with  $P_k$ . Suppose the  $i$ -th element of  $\overline{u_{k,t1}}$  is a variable (or constant). Then, the  $i$ -th element of  $\overline{u_{k,t2}}$  must also be a variable (or constant).
- (2) Consider any two rules  $r_{t1}$  and  $r_{t2}$  in the clique, and any terms  $\overline{u_{k1,t1}}$ ,  $\overline{u_{k1,t2}}$ ,  $\overline{u_{k2,t1}}$ ,  $\overline{u_{k2,t2}}$ , associated with any two predicates  $P_{k1}$  and  $P_{k2}$ , in these rules, respectively. Suppose the  $i$ -th element of  $\overline{u_{k1,t1}}$  and the  $j$ -th element of  $\overline{u_{k1,t2}}$  are common variables, i.e., they correspond to a join condition. Then the  $i$ -th element of  $\overline{u_{k2,t1}}$  and the  $j$ -th element of  $\overline{u_{k2,t2}}$  must also correspond to a join condition.
- (3) All the variables in the vector of terms associated with the predicates in the maximal set of the clique,  $P_1, \dots, P_p$ , must be mutually disjoint with the all the variables that occur in the vector of terms associated with the other predicates, in the other rules of the clique.

### 11.2. The Translation for the First-Order Predicates

#### Step 1

Each r-update rule  $q$  relevant to  $P(\overline{v})$  transforms each a-update rule  $p$  which is relevant to  $P(\overline{u})$ , or the fact  $P(\overline{u})$ , if it occurs in  $EDB_{init}$ . Two special unused predicates  $P^*$  and  $P^{**}$  are associated with each predicate  $P$  in the UR program which is transformed in this step.

Let the a-update rule  $p$  be of the following form:

$$p: A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow \text{assert } P(\overline{u})$$

where  $a$  or  $b$  could be equal to 0, i.e.,  $P(\overline{u})$  could be a fact.

Let each r-update rule  $q$  relevant to  $P$  be as follows:

$$q: P(\overline{v}), C_1, \dots, C_c, \neg D_1, \dots, \neg D_d \rightarrow \text{retract } P(\overline{v})$$

Let  $\overline{u}$  and  $\overline{v}$  unify with the most general unifier  $\theta$ . Then, the following rules are placed in  $\overline{UR}$ :

$$C_1, \dots, C_c, \neg D_1, \dots, \neg D_d \rightarrow P^*(\overline{v})$$

$$\neg P^*(\overline{v}) \theta, A_1 \theta, \dots, A_a \theta, \neg B_1 \theta, \dots, \neg B_b \theta \rightarrow P(\overline{u}) \theta$$

<sup>1</sup> we use  $\overline{v}$  to represent the vector of terms  $(\overline{v}_1, \dots, \overline{v}_n)$ .

<sup>2</sup> The conditions we present are very simple and may eliminate several cliques. We can define more complicated conditions that accommodate more cliques, but do not present the details here.

$$\neg P^{**}(\bar{u}), A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow P(\bar{u})$$

$$(\bar{v} = \bar{u}) \rightarrow P^{**}(\bar{u})$$

### Step 2

Each fact in  $EDB_{init}$  which is not modified by any r-update rule, i.e., there are no r-update rules relevant to this fact is placed unchanged in  $\overline{UR}$ .

Each a-update rule p relevant to  $P(\bar{u})$  which is as follows:

$$p: A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow \mathbf{assert} P(\bar{u})$$

and which is not modified by any r-update rule will derive the following rule in  $\overline{UR}$ :

$$A_1, \dots, A_a, \neg B_1, \dots, \neg B_b \rightarrow P(\bar{u})$$