



# 19

## SCHEMA REFINEMENT AND NORMAL FORMS

- What problems are caused by redundantly storing information?
- What are functional dependencies?
- What are normal forms and what is their purpose?
- What are the benefits of BCNF and 3NF?
- What are the considerations in decomposing relations into appropriate normal forms?
- Where does normalization fit in the process of database design?
- Are there general dependencies useful in database design?
- Key concepts: redundancy, insert, delete, and update anomalies; functional dependency, Armstrong's Axioms; dependency closure, attribute closure; normal forms, BCNF, 3NF; decompositions, lossless-join, dependency-preservation; multivalued dependencies, join dependencies, inclusion dependencies, 4NF, 5NF

It is a melancholy truth that even great men have their poor relations.

Charles Dickens

Conceptual database design gives us a set of relation schemas and integrity constraints (ICs) that can be regarded as a good starting point for the final database design. This initial design must be refined by taking the ICs into account more fully than is possible with just the ER model constructs and also by considering performance criteria and typical workloads. In this chapter, we discuss how ICs can be used to refine the conceptual schema produced by

translating an ER model design into a collection of relations. Workload and performance considerations are discussed in Chapter 20.

We concentrate on an important class of constraints called *functional dependencies*. Other kinds of les, for example, *multivalued dependencies* and *join dependencies*, also provide useful information. They can sometimes reveal redundancies that cannot be detected using functional dependencies alone. We discuss these other constraints briefly.

This chapter is organized as follows. Section 19.1 is an overview of the schema refinement approach discussed in this chapter. We introduce functional dependencies in Section 19.2. In Section 19.3, we show how to reason with functional dependency information to infer additional dependencies from a given set of dependencies. We introduce normal forms for relations in Section 19.4; the normal form satisfied by a relation is a measure of the redundancy in the relation. A relation with redundancy can be refined by *decomposing it*, or replacing it with smaller relations that contain the same information but without redundancy. We discuss decompositions and desirable properties of decompositions in Section 19.5, and we show how relations can be decomposed into smaller relations in desirable normal forms in Section 19.6.

In Section 19.7, we present several examples that illustrate how relational schemas obtained by translating an ER model design can nonetheless suffer from redundancy, and we discuss how to refine such schemas to eliminate the problems. In Section 19.8, we describe other kinds of dependencies for database design. We conclude with a discussion of normalization for our case study, the Internet shop, in Section 19.9.

## 19.1 INTRODUCTION TO SCHEMA REFINEMENT

We now present an overview of the problems that schema refinement is intended to address and a refinement approach based on decompositions. Redundant storage of information is the root cause of these problems. Although decomposition can eliminate redundancy, it can lead to problems of its own and should be used with caution.

### 19.1.1 Problems Caused by Redundancy

Storing the same information redundantly, that is, in more than one place within a database, can lead to several problems:

- **Redundant Storage:** Some information is stored repeatedly.

- ii **Update Anomalies:** If one copy of such repeated data is updated, an inconsistency is created unless all copies are similarly updated.
- ii **Insertion Anomalies:** It may not be possible to store certain information unless some other, unrelated, information is stored as well.
- ii **Deletion Anomalies:** It may not be possible to delete certain information without losing some other, unrelated, information as well.

Consider a relation obtained by translating a variant of the Hourly\_Emps entity set from Chapter 2:

Hourly\_Emps(*ssn*, *name*, *lot*, *rating*, *hourly\_wages*, *hours\_worked*)

In this chapter, we omit attribute type information for brevity, since our focus is on the grouping of attributes into relations. We often abbreviate an attribute name to a single letter and refer to a relation schema by a string of letters, one per attribute. For example, we refer to the Hourly\_Emps schema as *SNLRWH* (*W* denotes the *hourly\_wages* attribute).

The key for Hourly\_Emps is *ssn*. In addition, suppose that the *hourly\_wages* attribute is determined by the *rating* attribute. That is, for a given *rating* value, there is only one permissible *hourly\_wages* value. This IC is an example of a *functional dependency*. It leads to possible redundancy in the relation Hourly\_Emps, as illustrated in Figure 19.1.

	<i>name</i>	<i>lot</i>	<i>rating</i>	<i>hourly_wages</i>	<i>hours_worked</i>
123-22-3666	Attishoo	48	8	10	40
231-31-5368	Sruiley	22	8	10	30
131-24-3650	Srilethurst	35	5	7	30
434-26-3751	Guldu	35	5	7	32
612-67-4134	Madayan	35	8	10	40

Figure 19.1 An Instance of the Hourly\_Emps Relation

If the same value appears in the *rating* column of two tuples, the IC tells us that the same value must appear in the *hourly\_wages* column as well. This redundancy has the same negative consequences as before:

- ii **Redundant Storage:** the rating value 8 corresponds to the hourly wage 10, and this association is repeated three times.
- **Update Anomalies:** The *hourly\_wages* in the first tuple could be updated without making a similar change in the second tuple.

- *Insertion Anomalies:* We cannot insert a tuple for an employee unless we know the hourly wage for the employee's rating value.
- *Deletion Anomalies:* If we delete all tuples with a given rating value (e.g., we delete the tuples for Smithurst and Guldu) we lose the association between that *rating* value and its *hourly\_wage* value.

Ideally, we want schemas that do not permit redundancy, but at the very least we want to be able to identify schemas that do allow redundancy. Even if we choose to accept a schema with some of these drawbacks, perhaps owing to performance considerations, we want to make an informed decision.

## Null Values

It is worth considering whether the use of *null* values can address some of these problems. As we will see in the context of our example, they cannot provide a complete solution, but they can provide some help. In this chapter, we do not discuss the use of *null* values beyond this one example.

Consider the example *Hourly\_Emps* relation. Clearly, *null* values cannot help eliminate redundant storage or update anomalies. It appears that they can address insertion and deletion anomalies. For instance, to deal with the insertion anomaly example, we can insert an *employee* tuple with *null* values in the hourly wage field. However, *null* values cannot address all insertion anomalies. For example, we cannot record the hourly wage for a rating unless there is an employee with that rating, because we cannot store a null value in the *ssn* field, which is a primary key field. Similarly, to deal with the deletion anomaly example, we might consider storing a tuple with *null* values in all fields except *rating* and *hourly\_wages* if the last tuple with a given *rating* would otherwise be deleted. However, this solution does not work because it requires the *ssn* value to be *null*, and primary key fields cannot be *null*. Thus, *null* values do not provide a general solution to the problems of redundancy, even though they can help in some cases.

### 19.1.2 Decompositions

Intuitively, redundancy arises when a relational schema forces an association between attributes that is not natural. Functional dependencies (and, for that matter, other keys) can be used to identify such situations and suggest refinements to the schema. The essential idea is that many problems arising from redundancy can be addressed by replacing a relation with a collection of 'smaller' relations.

A decomposition of a relation schema  $R$  consists of replacing the relation schema by two (or more) relation schemas that each contain a subset of the attributes of  $R$  and together include all attributes in  $R$ . Intuitively, we want to store the information in any given instance of  $R$  by storing projections of the instance. This section examines the use of decompositions through several examples.

We can decompose Hourly\_Emps into two relations:

Hourly\_Emps2(ssn, name, lot, rating, hours\_worked)  
 Wages(rating, hourly\_wages)

The instances of these relations corresponding to the instance of Hourly\_Emps relation in Figure 19.1 is shown in Figure 19.2.

<u>ssn</u>	name	lot	rating	hours_worked
123-22-3666	Attishoo	48	8	40
231-31-5368	Sluiley	22	8	30
131-24-3650	Smethurst	35	5	30
434-26-3751	Guldu	35	5	32
612-67-4134	Madayan	35	8	40

rating	hourly_wages
8	10
5	7

Figure 19.2 Instances of Hourly\_Emps2 and Wages

Note that we can easily record the hourly wage for any rating simply by adding a tuple to Wages, even if no employee with that rating appears in the current instance of Hourly\_Emps. Changing the wage associated with a rating involves updating a single Wages tuple. This is more efficient than updating several tuples (as in the original design), and it eliminates the potential for inconsistency.

### 19.1.3 Problems Related to Decomposition

Unless we are careful, decomposing a relation schema can create more problems than it solves. Two important questions must be asked repeatedly:

1. Do we need to decompose a relation?

## 2. What problems (if any) does a given decomposition cause?

To help with the first question, several *normal forms* have been proposed for relations. If a relation schema is in one of these normal forms, we know that certain kinds of problems cannot arise. Considering the normal form of a given relation schema can help us to decide whether or not to decompose it further. If we decide that a relation schema must be decomposed further, we must choose a particular decomposition (i.e., a particular collection of smaller relations to replace the given relation).

With respect to the second question, two properties of decompositions are of particular interest. The *lossless-join* property enables us to recover any instance of the decomposed relation from corresponding instances of the smaller relations. The *dependency-preservation* property enables us to enforce any constraint on the original relation by simply enforcing the same constraints on each of the smaller relations. That is, we need not perform joins of the smaller relations to check whether a constraint on the original relation is violated.

From a performance standpoint, queries over the original relation may require us to join the decomposed relations. If such queries are common, the performance penalty of decomposing the relation may not be acceptable. In this case, we may choose to live with some of the problems of redundancy and not decompose the relation. It is important to be aware of the potential problems caused by such residual redundancy in the design and to take steps to avoid them (e.g., by adding some checks to application code). In some situations, decomposition could actually *improve* performance. This happens, for example, if most queries and updates examine only one of the decomposed relations, which is smaller than the original relation. We do not discuss the impact of decompositions on query performance in this chapter; this issue is covered in Section 20.8.

Our goal in this chapter is to explain some powerful concepts and design guidelines based on the theory of functional dependencies. A good database designer should have a firm grasp of normal forms and what problems they (do or do not) alleviate, the technique of decomposition, and potential problems with decompositions. For example, a designer often asks questions such as these: Is a relation in a given normal form? Is a decomposition dependency-preserving? Our objective is to explain when to raise these questions and the significance of the answers.

## 19.2 FUNCTIONAL DEPENDENCIES

A functional dependency (FD) is a kind of **lc** that generalizes the concept of a *key*. Let  $R$  be a relation schema and let  $X$  and  $Y$  be nonempty sets of attributes in  $R$ . We say that an instance  $r$  of  $R$  satisfies the  $\text{FD } X \rightarrow Y$ <sup>1</sup> if the following holds for every pair of tuples  $t_1$  and  $t_2$  in  $r$ :

If  $t_1.X = t_2.X$ , then  $t_1.Y = t_2.Y$ .

We use the notation  $t_1.X$  to refer to the projection of tuple  $t_1$  onto the attributes in  $X$ , in a natural extension of our TRC notation (see Chapter 4)  $t.a$  for referring to attribute  $a$  of tuple  $t$ . An  $\text{FD } X \rightarrow Y$  essentially says that if two tuples agree on the values in attributes  $X$ , they must also agree on the values in attributes  $Y$ .

Figure 19.3 illustrates the meaning of the  $\text{FD } AB \rightarrow C$  by showing an instance that satisfies this dependency. The first two tuples show that an FD is not the same as a key constraint: Although the FD is not violated,  $AB$  is clearly not a key for the relation. The third and fourth tuples illustrate that if two tuples differ in either the  $A$  field or the  $B$  field, they can differ in the  $C$  field without violating the FD. On the other hand, if we add a tuple  $(a_1, b_1, c_2, d_1)$  to the instance shown in this figure, the resulting instance would violate the FD; to see this violation, compare the first tuple in the figure with the new tuple.

$A$	$B$	$C$	$D$
a1	b1	c1	d1
a1	b1	c1	d2
a1	b2	c2	d1
a2	b1	c3	d1

Figure 19.3 An Instance that Satisfies  $AB \rightarrow C$

Recall that a *legal* instance of a relation must satisfy all specified **lc**s, including all specified FDs. As noted in Section 3.2, **lc**s must be identified and specified based on the semantics of the real-world enterprise being modeled. By looking at an instance of a relation, we might be able to tell that a certain FD does *not* hold. However, we can never deduce that an FD *does* hold by looking at one or more instances of the relation, because an FD, like other **lc**s, is a statement about *all* possible legal instances of the relation.

<sup>1</sup> $X \rightarrow Y$  is read as  $X$  *functionally determines*  $Y$ , or simply as  $X$  *determines*  $Y$ .

A primary key constraint is a special case of an FD. The attributes in the key play the role of  $X$ , and the set of all attributes in the relation plays the role of  $Y$ . Note, however, that the definition of an FD does not require that the set  $X$  be minimal; the additional minimality condition must be met for  $X$  to be a key. If  $X \rightarrow Y$  holds, where  $Y$  is the set of all attributes, and there is some (strictly contained) subset  $V$  of  $X$  such that  $V \rightarrow Y$  holds, then  $X$  is a *superkey*.

In the rest of this chapter, we see several examples of FDs that are not key constraints.

### 19.3 REASONING ABOUT FDS

Given a set of FDs over a relation schema  $R$ , typically several additional FDs hold over  $R$  whenever all of the given FDs hold. As an example, consider:

Workers(*ssn*, *name*, *lot*, *did*, *since*)

We know that  $ssn \rightarrow did$  holds, since *ssn* is the key, and FD  $did \rightarrow lot$  is given to hold. Therefore, in any legal instance of *Workers*, if two tuples have the same *ssn* value, they must have the same *did* value (from the first FD), and because they have the same *did* value, they must also have the same *lot* value (from the second FD). Therefore, the FD  $ssn \rightarrow lot$  also holds on *Workers*.

We say that an FD  $f$  is implied by a given set  $F$  of FDs if  $f$  holds on every relation instance that satisfies all dependencies in  $F$ ; that is,  $f$  holds whenever all FDs in  $F$  hold. Note that it is not sufficient for  $f$  to hold on some instance that satisfies all dependencies in  $F$ ; rather,  $f$  must hold on *every* instance that satisfies all dependencies in  $F$ .

#### 19.3.1 Closure of a Set of FDs

The set of all FDs implied by a given set  $F$  of FDs is called the closure of  $F$ , denoted as  $F^+$ . An important question is how we can infer, or compute, the closure of a given set  $F$  of FDs. The answer is simple and elegant. The following three rules, called Armstrong's Axioms, can be applied repeatedly to infer all FDs implied by a set  $F$  of FDs. We use  $X$ ,  $Y$ , and  $Z$  to denote sets of attributes over a relation schema  $R$ :

- Reflexivity: If  $X \supseteq Y$ , then  $X \rightarrow Y$ .
- Augmentation: If  $X \rightarrow Y$ , then  $XZ \rightarrow YZ$  for any  $Z$ .
- Transitivity: If  $X \rightarrow Y$  and  $Y \rightarrow Z$ , then  $X \rightarrow Z$ .



**Theorem 1** *Armstrong's Axioms are **sound**, in that they generate only FDs in  $F^+$  when applied to a set  $F$  of FDs. They are also **complete**, in that repeated application of these rules will generate all FDs in the closure  $F^+$ .*

The soundness of Armstrong's Axioms is straightforward to prove. Completeness is harder to show; see Exercise 19.17.

It is convenient to use some additional rules while reasoning about  $F^+$ :

- **Union:** If  $X \rightarrow Y$  and  $X \rightarrow Z$ , then  $X \rightarrow YZ$ .
- **Decomposition:** If  $X \rightarrow YZ$ , then  $X \rightarrow Y$  and  $X \rightarrow Z$ .

These additional rules are not essential; their soundness can be proved using Armstrong's Axioms.

To illustrate the use of these inference rules for FDs, consider a relation schema  $ABC$  with FDs  $A \rightarrow B$  and  $B \rightarrow C$ . In a trivial FD, the right side contains only attributes that also appear on the left side; such dependencies always hold due to reflexivity. Using reflexivity, we can generate all trivial dependencies, which are of the form:

$$X \rightarrow Y, \text{ where } Y \subseteq X, X \subseteq ABC, \text{ and } Y \subseteq ABC.$$

From transitivity we get  $A \rightarrow C$ . From augmentation we get the nontrivial dependencies:

$$AC \rightarrow BC, AB \rightarrow AC, AB \rightarrow C.$$

As another example, we use a more elaborate version of Contracts:

$$\text{Contracts}(\text{contractid}, \text{supplierid}, \text{projectid}, \text{deptid}, \text{partid}, \text{qty}, \text{value})$$

We denote the schema for Contracts as  $CSJDPQV$ . The meaning of a tuple is that the contract with *contractid*  $C$  is an agreement that supplier  $S(\text{supplierid})$  will supply  $Q$  items of *part*? (*partid*) to project  $J$  (*projectid*) associated with department  $D$  (*deptid*); the value  $V$  of this contract is equal to *value*.

The following facts are known to hold:

1. The contract id  $C$  is a key:  $C \rightarrow CSJDPQV$ .
2. A project purchases a given part using a single contract:  $(J, P) \rightarrow C$ .

3. A department purchases at most one part from a supplier:  $SD \rightarrow P$ .

Several additional FDs hold in the closure of the set of given FDs:

From  $JP \rightarrow C$ ,  $C \rightarrow CSJDPQV$ , and transitivity, we infer  $JP \rightarrow CSJDPQV$ .

From  $SD \rightarrow P$  and augmentation, we infer  $SDJ \rightarrow JP$ .

From  $SDJ \rightarrow JP$ ,  $JP \rightarrow CSJDPQV$ , and transitivity, we infer  $SDJ \rightarrow CSJDPQV$ . (Incidentally, while it may appear tempting to do so, we *cannot* conclude  $SD \rightarrow CSJDPQV$ , canceling  $J$  on both sides. FD inference is not like arithmetic multiplication!)

We can infer several additional FDs that are in the closure by using augmentation or decomposition. For example, from  $C \rightarrow CSJDPQV$ , using decomposition, we can infer:

$C \rightarrow C$ ,  $C \rightarrow S$ ,  $C \rightarrow J$ ,  $C \rightarrow D$ , and so forth

Finally, we have a number of trivial FDs from the reflexivity rule.

### 19.3.2 Attribute Closure

If we just want to check whether a given dependency, say,  $X \rightarrow Y$ , is in the closure of a set  $F$  of FDs, we can do so efficiently without computing  $F^+$ . We first compute the attribute closure  $X^+$  with respect to  $F$ , which is the set of attributes  $A$  such that  $X \rightarrow A$  can be inferred using the Armstrong Axioms. The algorithm for computing the attribute closure of a set  $X$  of attributes is shown in Figure 19.4.

```

closure = X;
repeat until there is no change: {
    if there is an FD  $U \rightarrow V$  in  $F$  such that  $U \subseteq \text{closure}$ ,
        then set  $\text{closure} = \text{closure} \cup V$ 
}
```

Figure 19.4 Computing the Attribute Closure of Attribute Set  $X$

**Theorem 2** The algorithm shown in Figure 19.4 computes the attribute closure  $X^+$  of the attribute set  $X$  with respect to the set of FDs  $F$ .

The proof of this theorem is considered in Exercise 19.15. This algorithm can be modified to find keys by starting with set  $X$  containing a single attribute and stopping as soon as *closure* contains all attributes in the relation schema. By varying the starting attribute and the order in which the algorithm considers FDs, we can obtain all candidate keys.

## 19.4 NORMAL FORMS

Given a relation schema, we need to decide whether it is a good design or we need to decompose it into smaller relations. Such a decision must be guided by an understanding of what problems, if any, arise from the current schema. To provide such guidance, several normal forms have been proposed. If a relation schema is in one of these normal forms, we know that certain kinds of problems cannot arise.

The normal forms based on FDs are *first normal form (1NF)*, *second normal form (2NF)*, *third normal form (3NF)*, and *Boyce-Codd normal form (BCNF)*. These forms have increasingly restrictive requirements: Every relation in BCNF is also in 3NF, every relation in 3NF is also in 2NF, and every relation in 2NF is in 1NF. A relation is in first normal form if every field contains only atomic values, that is, no lists or sets. This requirement is implicit in our definition of the relational model. Although some of the newer database systems are relaxing this requirement, in this chapter we assume that it always holds. 2NF is mainly of historical interest. 3NF and BCNF are important from a database design standpoint.

While studying normal forms, it is important to appreciate the role played by FDs. Consider a relation schema  $R$  with attributes  $ABC$ . In the absence of any ICs, any set of ternary tuples is a legal instance and there is no potential for redundancy. On the other hand, suppose that we have the FD  $A \rightarrow B$ . Now if several tuples have the same  $A$  value, they must also have the same  $B$  value. This potential redundancy can be predicted using the FD information. If more detailed ICs are specified, we may be able to detect more subtle redundancies as well.

We primarily discuss redundancy revealed by FD information. In Section 19.8, we discuss more sophisticated ICs called *multivalued dependencies* and *join dependencies* and normal forms based on them.

### 19.4.1 Boyce-Codd Normal Form

Let  $R$  be a relation schema,  $F$  be the set of FDs given to hold over  $R$ ,  $X$  be a subset of the attributes of  $R$ , and  $A$  be an attribute of  $R$ .  $R$  is in Boyce-Codd

normal form if, for every  $F1)X \rightarrow A$  in  $F$ , one of the following statements is true:

- $A \in X$ ; that is, it is a trivial FD, or
- $X$  is a superkey.

Intuitively, in a BCNF relation, the only nontrivial dependencies are those in which a key determines some attribute(s). Therefore, each tuple can be thought of as an entity or relationship, identified by a key and described by the remaining attributes. Rent (in [425]) puts this colorfully, if a little loosely: "Each attribute must describe [an entity or relationship identified by] the key, the whole key, and nothing but the key." If we use ovals to denote attributes or sets of attributes and draw arcs to indicate FDs, a relation in BCNF has the structure illustrated in Figure 19.5, considering just one key for simplicity. (If there are several candidate keys, each candidate key can play the role of KEY in the figure, with the other attributes being the ones not in the chosen candidate key.)

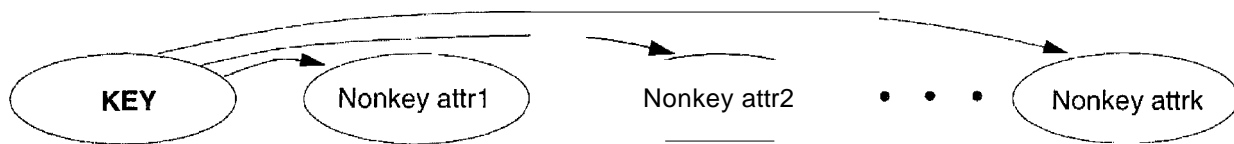


Figure 19.5 FDs in a BCNF Relation

BCNF ensures that no redundancy can be detected using FD information alone. It is thus the most desirable normal form (from the point of view of redundancy) if we take into account only FD information. This point is illustrated in Figure 19.6.

$X$	$Y$	$A$
$x$	$y_1$	$a$
$x$	$y_2$	?

Figure 19.6 Instance Illustrating BCNF

This figure shows (two tuples in) an instance of a relation with three attributes  $X$ ,  $Y$ , and  $A$ . There are two tuples with the same value in the  $X$  column. Now suppose that we know that this instance satisfies an FD  $X \rightarrow A$ . We can see that one of the tuples has the value  $a$  in the  $A$  column. What can we infer about the value in the  $A$  column in the second tuple? Using the FD, we can conclude that the second tuple also has the value  $a$  in this column. (Note that this is really the only kind of inference we can make about values in the fields of tuples by using FDs.)

But is this situation not an example of redundancy? We appear to have stored the value  $a$  twice. Can such a situation arise in a BCNF relation? The answer is No! If this relation is in BCNF, because  $A$  is distinct from  $X$ , it follows that  $X$  must be a key. (Otherwise, the FD  $X \rightarrow A$  would violate BCNF.) If  $X$  is a key, then  $Y_1 = Y_2$ , which means that the two tuples are identical. Since a relation is defined to be a *set* of tuples, we cannot have two copies of the same tuple and the situation shown in Figure 19.6 cannot arise.

Therefore, if a relation is in BCNF, every field of every tuple records a piece of information that cannot be inferred (using only FDs) from the values in all other fields in (all tuples of) the relation instance.

### 19.4.2 Third Normal Form

Let  $R$  be a relation schema,  $F$  be the set of FDs given to hold over  $R$ ,  $X$  be a subset of the attributes of  $R$ , and  $A$  be an attribute of  $R$ .  $R$  is in third normal form if, for every FD  $X \rightarrow A$  in  $F$ , one of the following statements is true:

- $A \in X$ ; that is, it is a trivial FD, or
- $X$  is a superkey, or
- $A$  is part of some key for  $R$ .

The definition of 3NF is similar to that of BCNF, with the only difference being the third condition. Every BCNF relation is also in 3NF. To understand the third condition, recall that a key for a relation is a *minimal* set of attributes that uniquely determines all other attributes.  $A$  must be part of a key (any key, if there are several). It is not enough for  $A$  to be part of a superkey, because the latter condition is satisfied by every attribute! Finding all keys of a relation schema is known to be an NP-complete problem, and so is the problem of determining whether a relation schema is in 3NF.

Suppose that a dependency  $X \rightarrow A$  causes a violation of 3NF. There are two cases:

- $X$  is a *proper subset of some key*  $K$ . Such a dependency is sometimes called a **partial dependency**. In this case, we store  $(X, A)$  pairs redundantly. As an example, consider the Reserves relation with attributes  $SBDC$  from Section 19.7.4. The only key is  $SEI$ , and we have the FD  $S \rightarrow C$ . We store the credit card number for a sailor as many times as there are reservations for that sailor.
- $X$  is *not a proper subset of any key*. Such a dependency is sometimes called a **transitive dependency**, because it means we have a chain of

dependencies  $X \rightarrow A$ . The problem is that we cannot associate an  $X$  value with a  $K$  value unless we also associate an  $A$  value with an  $X$  value. As an example, consider the *Hourly\_Emps* relation with attributes *SNLRWH* from Section 19.7.1. The only key is  $S$ , but there is an FD  $R \rightarrow W$ , which gives rise to the chain  $S \rightarrow R \rightarrow W$ . The consequence is that we cannot record the fact that employee  $S$  has rating  $R$  without knowing the hourly wage for that rating. This condition leads to insertion, deletion, and update anomalies.

Partial dependencies are illustrated in Figure 19.7, and transitive dependencies are illustrated in Figure 19.8. Note that in Figure 19.8, the set  $X$  of attributes may or may not have some attributes in common with  $KEY$ ; the diagram should be interpreted as indicating only that  $X$  is not a subset of  $KEY$ .

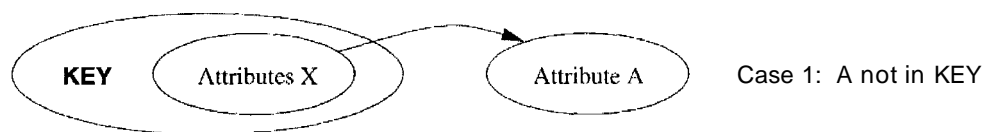


Figure 19.7 Partial Dependencies

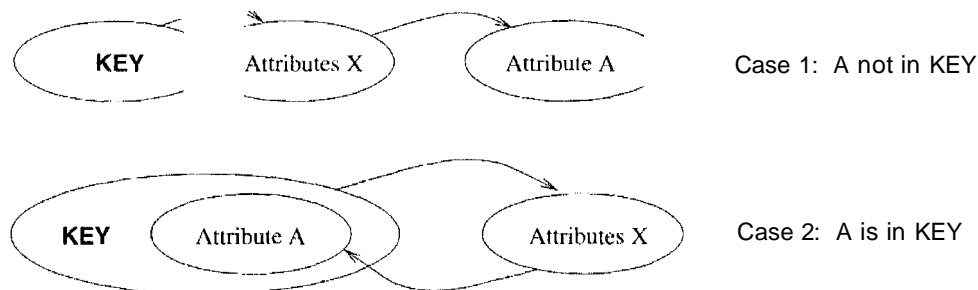


Figure 19.8 Transitive Dependencies

The motivation for 3NF is rather technical. By making an exception for certain dependencies involving key attributes, we can ensure that every relation schema can be decomposed into a collection of 3NF relations using only decompositions that have certain desirable properties (Section 19.5). Such a guarantee does not exist for BCNF relations; the 3NF definition weakens the BCNF requirements just enough to make this guarantee possible. We may therefore compromise by settling for a 3NF design. As we see in Chapter 20, we may sometimes accept this compromise (or even settle for a non-3NF schema) for other reasons as well.

Unlike BCNF, however, some redundancy is possible with 3NF. The problems associated with partial and transitive dependencies persist if there is a nontrivial dependency  $X \rightarrow A$  and  $X$  is not a superkey, even if the relation is in 3NF because  $A$  is part of a key. To understand this point, let us revisit the *Reserves*

relation with attributes *SEDe* and the FD  $S \rightarrow C$ , which states that a sailor uses a unique credit card to pay for reservations.  $S$  is not a key, and  $C$  is not part of a key. (In fact, the only key is *SED*.) Hence, this relation is not in 3NF;  $(S, C)$  pairs are stored redundantly. However, if we also know that credit cards uniquely identify the owner, we have the FD  $C \rightarrow S$ , which means that *CBD* is also a key for Reserves. Therefore, the dependency  $S \rightarrow C$  does not violate 3NF, and Reserves is in 3NF. Nonetheless, in all tuples containing the same  $S$  value, the same  $(S, C)$  pair is redundantly recorded.

For completeness, we remark that the definition of second normal form is essentially that partial dependencies are not allowed. Thus, if a relation is in 3NF (which precludes both partial and transitive dependencies), it is also in 2NF.

## 19.5 PROPERTIES OF DECOMPOSITIONS

Decomposition is a tool that allows us to eliminate redundancy. As noted in Section 19.1.3, however, it is important to check that a decomposition does not introduce new problems. In particular, we should check whether a decomposition allows us to recover the original relation, and whether it allows us to check integrity constraints efficiently. We discuss these properties next.

### 19.5.1 Lossless-Join Decomposition

Let  $R$  be a relation schema and let  $F$  be a set of FDs over  $R$ . A decomposition of  $R$  into two schemas with attribute sets  $X$  and  $Y$  is said to be a lossless-join decomposition with respect to  $F$  if, for every instance  $r$  of  $R$  that satisfies the dependencies in  $F$ ,  $\pi_X(r) \bowtie \pi_Y(r) = r$ . In other words, we can recover the original relation from the decomposed relations.

This definition can easily be extended to cover a decomposition of  $R$  into more than two relations. It is easy to see that  $r \subseteq \pi_X(r) \bowtie \pi_Y(r)$  always holds. In general, though, the other direction does not hold. If we take projections of a relation and recombine them using natural join, we typically obtain some tuples that were not in the original relation. This situation is illustrated in Figure 19.9.

By replacing the instance  $r$  shown in Figure 19.9 with the instances  $\pi_{SP}(r)$  and  $\pi_{PI}(r)$ , we lose some information. In particular, suppose that the tuples in  $r$  denote relationships. We can no longer tell that the relationships  $(s_1, p_1, d_3)$  and  $(s_3, p_1, d_1)$  do not hold. The decomposition of schema *SPD* into *SP* and *PI* is therefore lossy if the instance  $r$  shown in the figure is legal, that is, if this

<i>S</i>	<i>P</i>	<i>D</i>		<i>S</i>	<i>P</i>		<i>P</i>	<i>D</i>		<i>S</i>	<i>P</i>	<i>D</i>
s1	pl.	ell		s1	pI		p1	d1		81	pI	ell
s2	p2	d2		s2	p2		p2	d2		82	p2	d2
s3	pl.	d3		s3	pI		pI	d3		83	pI	d3
										s1	pI	d3
										s3	pI	ell

Instance *r*                       $\pi_{SP}(r)$                        $\pi_{PD}(r)$                        $\pi_{SP}(r) \bowtie \pi_{PD}(r)$

Figure 19.9 Instances Illustrating Lossy Decompositions

instance could arise in the enterprise being modeled. (Observe the similarities between this example and the Contracts relationship set in Section 2.5.3.)

*All decompositions used to eliminate redundancy must be lossless.* The following simple test is very useful:

**Theorem 3** *Let  $R$  be a relation and  $F$  be a set of FDs that hold over  $R$ . The decomposition of  $R$  into relations with attribute sets  $R_1$  and  $R_2$  is lossless if and only if  $F$  contains either the FD  $R_1 \twoheadrightarrow R_2$  or the FD  $R_2 \twoheadrightarrow R_1$ .*

In other words, the attributes common to  $R_1$  and  $R_2$  must contain a key for either  $R_1$  or  $R_2$ .<sup>2</sup> If a relation is decomposed into more than two relations, an efficient (time polynomial in the size of the dependency set) algorithm is available to test whether or not the decomposition is lossless, but we will not discuss it.

Consider the Hourly\_Emps relation again. It has attributes *SNLRWH*, and the FD  $R \twoheadrightarrow W$  causes a violation of 3NF. We dealt with this violation by decomposing the relation into *SNLRH* and *RW*. Since  $R$  is common to both decomposed relations and  $R \twoheadrightarrow W$  holds, this decomposition is lossless-join.

This example illustrates a general observation that follows from Theorem 3:

If an FD  $X \twoheadrightarrow Y$  holds over a relation  $R$  and  $X \cap Y$  is empty, the decomposition of  $R$  into  $R - Y$  and  $XY$  is lossless.

$X$  appears in both  $R - Y$  (since  $X \cap Y$  is empty) and  $XY$ , and it is a key for  $XY$ .

<sup>2</sup>See Exercise 19.19 for a proof of Theorem 3. Exercise 19.11 illustrates that the 'only if' claim depends on the assumption that only functional dependencies can be specified as integrity constraints.



Another important observation, which we state without proof, has to do with repeated decomposition. Suppose that a relation  $R$  is decomposed into  $R_1$  and  $R_2$  through a lossless-join decomposition, and that  $R_1$  is decomposed into  $R_{11}$  and  $R_{12}$  through another lossless-join decomposition. Then, the decomposition of  $R$  into  $R_{11}$ ,  $R_{12}$ , and  $R_2$  is lossless-join; by joining  $R_{11}$  and  $R_{12}$ , we can recover  $R_1$ , and by then joining  $R_1$  and  $R_2$ , we can recover  $R$ .

### 19.5.2 Dependency-Preserving Decomposition

Consider the *Contracts* relation with attributes  $C, S, J, D, P, C, J, V$  from Section 19.3.1. The given FDs are  $C \rightarrow CSJDPQV$ ,  $JP \rightarrow C$ , and  $SD \rightarrow P$ . Because  $SD$  is not a key the dependency  $SD \rightarrow P$  causes a violation of BCNF.

We can decompose *Contracts* into two relations with schemas  $CSJDQV$  and  $SDP$  to address this violation; the decomposition is lossless-join. There is one subtle problem, however. We can enforce the integrity constraint  $JP \rightarrow C$  easily when a tuple is inserted into *Contracts* by ensuring that no existing tuple has the same  $JP$  values (as the inserted tuple) but different  $C$  values. Once we decompose *Contracts* into  $CSJDQV$  and  $SDP$ , enforcing this constraint requires an expensive join of the two relations whenever a tuple is inserted into  $CSJDQV$ . We say that this decomposition is not dependency-preserving.

Intuitively, a *dependency-preserving decomposition* allows us to enforce all FDs by examining a single relation instance on each insertion or modification of a tuple. (Note that deletions cannot cause violation of FDs.) To define dependency-preserving decompositions precisely, we have to introduce the concept of a projection of FDs.

Let  $R$  be a relation schema that is decomposed into two schemas with attribute sets  $X$  and  $Y$ , and let  $F$  be a set of FDs over  $R$ . The **projection of  $F$  on  $X$**  is the set of FDs in the closure  $F^+$  (not just  $F$ !) that involve only attributes in  $X$ . We denote the projection of  $F$  on attributes  $X$  as  $F_X$ . Note that a dependency  $U \rightarrow V$  in  $F^+$  is in  $F_X$  only if *all* the attributes in  $U$  and  $V$  are in  $X$ .

The decomposition of relation schema  $R$  with FDs  $F$  into schemas with attribute sets  $X$  and  $Y$  is dependency-preserving if  $(F_X \cup F_Y)^+ = F^+$ . That is, if we take the dependencies in  $F_X$  and  $F_Y$  and compute the closure of their union, we get back all dependencies in the closure of  $F$ . Therefore, we need to enforce only the dependencies in  $F_X$  and  $F_Y$ ; all FDs in  $F^+$  are then sure to be satisfied. To enforce  $F_X$ , we need to examine only relation  $X$  (on inserts to that relation). To enforce  $F_Y$ , we need to examine only relation  $Y$ .

To appreciate the need to consider the closure  $F^+$  while considering the projection of  $F$ , suppose that a relation  $R$  with attributes  $ABC$  is decomposed into relations with attributes  $AB$  and  $BC$ . The set  $F$  of FDs over  $R$  includes  $A \rightarrow B$ ,  $B \rightarrow C$ , and  $C \rightarrow A$ . Of these,  $A \rightarrow B$  is in  $F_{AB}$  and  $B \rightarrow C$  is in  $F_{BC}$ . But is this decomposition dependency-preserving? What about  $C \rightarrow A$ ? This dependency is not implied by the dependencies listed (thus far) for  $F_{AB}$  and  $F_{BC}$ .

The closure of  $F$  contains all dependencies in  $F$  plus  $A \rightarrow C$ ,  $B \rightarrow A$ , and  $C \rightarrow B$ . Consequently,  $F_{AB}$  also contains  $B \rightarrow A$ , and  $F_{BC}$  contains  $C \rightarrow B$ . Therefore,  $F_{AB} \cup F_{BC}$  contains  $A \rightarrow B$ ,  $B \rightarrow C$ ,  $B \rightarrow A$ , and  $C \rightarrow B$ . The closure of the dependencies in  $F_{AB}$  and  $F_{BC}$  now includes  $C \rightarrow A$  (which follows from  $C \rightarrow B$ ,  $B \rightarrow A$ , and transitivity). Thus, the decomposition preserves the dependency  $C \rightarrow A$ .

A direct application of the definition gives us a straightforward algorithm for testing whether a decomposition is dependency-preserving. (This algorithm is exponential in the size of the dependency set. A polynomial algorithm is available; see Exercise 19.9.)

We began this section with an example of a lossless-join decomposition that was not dependency-preserving. Other decompositions are dependency-preserving, but not lossless. A simple example consists of a relation  $ABC$  with FD  $A \rightarrow B$  that is decomposed into  $AB$  and  $BC$ .

## 19.6 NORMALIZATION

Having covered the concepts needed to understand the role of normal forms and decompositions in database design, we now consider algorithms for converting relations to BCNF or 3NF. If a relation schema is not in BCNF, it is possible to obtain a lossless-join decomposition into a collection of BCNF relation schemas. Unfortunately, there may be no dependency-preserving decomposition into a collection of BCNF relation schemas. However, there is always a dependency-preserving, lossless-join decomposition into a collection of 3NF relation schemas.

### 19.6.1 Decomposition into BCNF

We now present an algorithm for decomposing a relation schema  $R$  with a set of FDs into a collection of BCNF relation schemas:

1. Suppose that  $R$  is not in BCNF. Let  $X \subset R$ ,  $A$  be a single attribute in  $R$ , and  $X \rightarrow A$  be an FD that causes a violation of BCNF. Decompose  $R$  into  $R - A$  and  $XA$ .
2. If either  $R - A$  or  $XA$  is not in BCNF, decompose them further by a recursive application of this algorithm.

$R - A$  denotes the set of attributes other than  $A$  in  $R$ , and  $XA$  denotes the union of attributes in  $X$  and  $A$ . Since  $X \rightarrow A$  violates BCNF, it is not a trivial dependency; further,  $A$  is a single attribute. Therefore,  $A$  is not in  $X$ ; that is,  $X \cap A$  is empty. Therefore, each decomposition carried out in Step 1 is lossless-join.

The set of dependencies associated with  $R - A$  and  $XA$  is the projection of  $F$  onto their attributes. If one of the new relations is not in BCNF, we decompose it further in Step 2. Since a decomposition results in relations with strictly fewer attributes, this process terminates, leaving us with a collection of relations that are all in BCNF. Further, joining instances of the (two or more) relations obtained through this algorithm yields precisely the corresponding instance of the original relation (i.e., the decomposition into a collection of relations each of which is in BCNF is a lossless-join decomposition).

Consider the Contracts relation with attributes  $CSDJPQV$  and key  $C$ . We are given FDs  $JP \rightarrow C$  and  $3D \rightarrow P$ . By using the dependency  $3D \rightarrow P$  to guide the decomposition, we get the two schemas  $3DP$  and  $C5JDQV$ .  $3DP$  is in BCNF. Suppose that we also have the constraint that each project deals with a single supplier:  $J \rightarrow S$ . This means that the schema  $C5JDQV$  is not in BCNF. So we decompose it further into  $JS$  and  $CJDQV$ .  $C \rightarrow JDQV$  holds over  $CJDQV$ ; the only other FDs that hold are those obtained from this PI by augmentation, and therefore all FDs contain a key in the left side. Thus, each of the schemas  $3DP$ ,  $JS$ , and  $CJDQV$  is in BCNF, and this collection of schemas also represents a lossless-join decomposition of  $C5JDQV$ .

The steps in this decomposition process can be visualized as a tree, as shown in Figure 19.10. The root is the original relation  $C5JJPQV$ , and the leaves are the BCNF relations that result from the decomposition algorithm:  $3DP$ ,  $JS$ , and  $CJDQV$ . Intuitively, each internal node is replaced by its children through a single decomposition step guided by the FD shown just below the node.

## Redundancy in BCNF Revisited

The decomposition of  $C5JDQV$  into  $3DP$ ,  $JS$ , and  $CJDQV$  is not dependency-preserving. Intuitively, dependency  $JP \rightarrow C$  cannot be enforced without a join. One way to deal with this situation is to add a relation with attributes  $GP$ . In

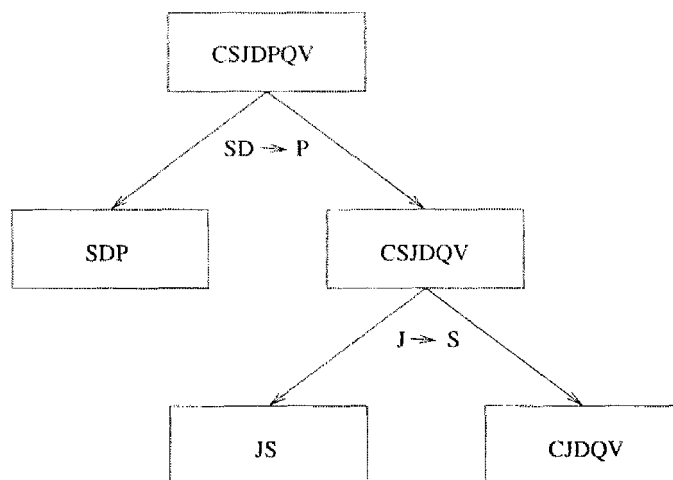


Figure 19.10 Decomposition of *CSJDPQV* into *SDP*, *JS*, and *CJDQV*

effect, this solution amounts to storing some information redundantly to make the dependency enforcement cheaper.

This is a subtle point: Each of the schemas *CJP*, *SDP*, *JS*, and *CJDQV* is in BCNF, yet some redundancy can be predicted by FD information. In particular, if we join the relation instances for *SDP* and *CJDQV* and project the result onto the attributes *CJP*, we must get exactly the instance stored in the relation with schema *CJP*. We saw in Section 19.4.1 that there is no such redundancy within a single BCNF relation. This example shows that redundancy can still occur across relations, even though there is no redundancy within a relation.

## Alternatives in Decomposing to BCNF

Suppose several dependencies violate BCNF. Depending on which of these dependencies we choose to guide the next decomposition step, we may arrive at quite different collections of BCNF relations. Consider *Contracts*. We just decomposed it into *SDP*, *JS*, and *CJDQV*. Suppose we choose to decompose the original relation *CSJDPQV* into *JS* and *CJDPQV*, based on the FD  $I \rightarrow S$ . The only dependencies that hold over *CJDPQV* are  $IP \rightarrow C$  and the key dependency  $C \rightarrow C.IDPQV$ . Since *IP* is a key, *CJDPQV* is in BCNF. Thus, the schemas *JS* and *CJDPQV* represent a lossless-join decomposition of *Contracts* into BCNF relations.

The lesson to be learned here is that the theory of dependencies can tell us when there is redundancy and give us clues about possible decompositions to address the problem, but it cannot discriminate among decomposition alternatives. A designer has to consider the alternatives and choose one based on the semantics of the application.

## BCNF and Dependency-Preservation

Sometimes, there simply is no decomposition into BCNF that is dependency-preserving. As an example, consider the relation schema  $SBD$ , in which a tuple denotes that sailor  $S$  has reserved boat  $B$  on date  $D$ . If we have the FDs  $SB \rightarrow D$  (a sailor can reserve a given boat for at most one day) and  $D \rightarrow B$  (on any given day at most one boat can be reserved),  $SBD$  is not in BCNF because  $D$  is not a key. If we try to decompose it, however, we cannot preserve the dependency  $SB \rightarrow D$ .

### 19.6.2 Decomposition into 3NF

Clearly, the approach we outlined for lossless-join decomposition into BCNF also gives us a lossless-join decomposition into 3NF. (Typically, we can stop a little earlier if we are satisfied with a collection of 3NF relations.) But this approach does not ensure dependency-preservation.

A simple modification, however, yields a decomposition into 3NF relations that is lossless-join and dependency-preserving. Before we describe this modification, we need to introduce the concept of a minimal cover for a set of FDs.

### Minimal Cover for a Set of FDs

A minimal cover for a set  $F$  of FDs is a set  $G$  of FDs such that:

1. Every dependency in  $G$  is of the form  $X \rightarrow A$ , where  $A$  is a single attribute.
2. The closure  $F^+$  is equal to the closure  $G^+$ .
3. If we obtain a set  $H$  of dependencies from  $G$  by deleting one or more dependencies or by deleting attributes from a dependency in  $G$ , then  $H^+ \neq F^+$ .

Intuitively, a minimal cover for a set  $F$  of FDs is an equivalent set of dependencies that is *minimal* in two respects: (1) Every dependency is as small as possible; that is, each attribute on the left side is necessary and the right side is a single attribute. (2) Every dependency in it is required for the closure to be equal to  $F^+$ .

As an example, let  $F$  be the set of dependencies:

$$A \rightarrow B, ABCD \rightarrow E, EF \rightarrow G, EF \rightarrow H, \text{ and } CDF \rightarrow EG.$$

First, let us rewrite  $CDF \rightarrow EG$  so that every right side is a single attribute:

$$ACDF \rightarrow E \text{ and } ACDF \rightarrow G,$$

Next consider  $ACDF \rightarrow G$ . This dependency is implied by the following FDs:

$$A \rightarrow B, ABCD \rightarrow E, \text{ and } EF \rightarrow G,$$

Therefore, we can delete it. Similarly, we can delete  $ACDF \rightarrow E$ . Next consider  $ABCD \rightarrow E$ . Since  $A \rightarrow B$  holds, we can replace it with  $ACD \rightarrow E$ . (At this point, the reader should verify that each remaining FD is minimal and required.) Thus, a minimal cover for  $F$  is the set:

$$A \rightarrow B, ACD \rightarrow E, EF \rightarrow G, \text{ and } EF \rightarrow H,$$

The preceding example illustrates a general algorithm for obtaining a minimal cover of a set  $F$  of FDs:

1. **Put the FDs in a Standard Form:** Obtain a collection  $G$  of equivalent FDs with a single attribute on the right side (using the decomposition axiom),
2. **Minimize the Left Side of Each FD:** For each FD in  $G$ , check each attribute in the left side to see if it can be deleted while preserving equivalence to  $F^+$ ,
3. **Delete Redundant FDs:** Check each remaining FD in  $G$  to see if it can be deleted while preserving equivalence to  $F^+$ ,

Note that the order in which we consider FDs while applying these steps could produce different minimal covers; there could be several minimal covers for a given set of FDs,

It's important, it is necessary to minimize the left sides of FDs *before* checking for redundant FDs. If these two steps are reversed, the final set of FDs could still contain some redundant FDs (i.e., not be a minimal cover), as the following example illustrates. Let  $F$  be the set of dependencies, each of which is already in the standard form:

$$ABCD \rightarrow E, E \rightarrow D, A \rightarrow B, \text{ and } AC \rightarrow D,$$

Observe that none of these FDs is redundant; if we checked for redundant FDs first, we would get the same set of FDs  $F$ . The left side of  $ABCD \rightarrow E$  can be replaced by  $AC$  while preserving equivalence to  $F^+$ , and we would stop here if we checked for redundant FDs in  $F$  before minimizing the left sides. However, the set of FDs we have is not a minimal cover:

$$AC \rightarrow E, E \twoheadrightarrow D, A \rightarrow B, \text{ and } AC \rightarrow D.$$

From transitivity, the first two FDs imply the last FD, which can therefore be deleted while preserving equivalence to  $F^+$ . The important point to note is that  $AC \rightarrow D$  becomes redundant only after we replace  $AB \twoheadrightarrow E$  with  $AC \rightarrow E$ . If we minimize left sides of FDs first and then check for redundant FDs, we are left with the first three FDs in the preceding list, which is indeed a minimal cover for  $F$ .

## Dependency-Preserving Decomposition into 3NF

Returning to the problem of obtaining a lossless-join, dependency-preserving decomposition into 3NF relations, let  $R$  be a relation with a set  $F$  of FDs that is a minimal cover, and let  $R_1, R_2, \dots, R_n$  be a lossless-join decomposition of  $R$ . For  $1 \leq i \leq n$ , suppose that each  $R_i$  is in 3NF and let  $F_i$  denote the projection of  $F$  onto the attributes of  $R_i$ . Do the following:

- Identify the set  $N$  of dependencies in  $F$  that is not preserved, that is, not included in the closure of the union of  $F_i$ s.
- For each FD  $X \rightarrow A$  in  $N$ , create a relation schema  $XA$  and add it to the decomposition of  $R$ .

Obviously, every dependency in  $F$  is preserved if we replace  $R$  by the  $R_i$ s plus the schemas of the form  $XA$  added in this step. The  $R_i$ s are given to be in 3NF. We can show that each of the schemas  $XA$  is in 3NF as follows: Since  $X \rightarrow A$  is in the minimal cover  $F$ ,  $Y \rightarrow A$  does not hold for any  $Y$  that is a strict subset of  $X$ . Therefore,  $X$  is a key for  $XA$ . Further, if any other dependencies hold over  $XA$ , the right side can involve only attributes in  $X$  because  $A$  is a single attribute (because  $X \rightarrow A$  is an FD in a minimal cover). Since  $X$  is a key for  $XA$ , none of these additional dependencies causes a violation of 3NF (although they might cause a violation of BCNF).

As an optimization, if the set  $N$  contains several FDs with the same left side, say,  $X \rightarrow A_1, X \rightarrow A_2, \dots, X \twoheadrightarrow A_n$ , we can replace them with a single equivalent FD  $X \rightarrow A_1 \dots A_n$ . Therefore, we produce one relation schema  $XA_1 \dots A_n$ , instead of several schemas  $XA_1, \dots, XA_n$ , which is generally preferable.

Consider the Contracts relation with attributes  $CSJDPQV$  and FDs  $JP \rightarrow C$ ,  $SD \rightarrow P$ , and  $J \twoheadrightarrow S$ . If we decompose  $CSJDPQV$  into  $SDIJ$  and  $CSJDQV$ , then  $SDP$  is in BCNF, but  $CSJDQV$  is not even in 3NF. So we decompose it further into  $JS$  and  $CJDQV$ . The relation schemas  $SDP$ ,  $JS$ , and  $CJDQV$  are in 3NF (in fact, in BCNF), and the decomposition is lossless-join. However,

the dependency  $JP \rightarrow C$  is not preserved. This problem can be addressed by adding a relation schema  $CJP$  to the decomposition.

## 3NF Synthesis

We assumed that the design process starts with an ER diagram, and that our use of FDs is primarily to guide decisions about decomposition. The algorithm for obtaining a lossless-join, dependency-preserving decomposition was presented in the previous section from this perspective-----a lossless-join decomposition into 3NF is straightforward, and the algorithm addresses dependency-preservation by adding extra relation schemas.

An alternative approach, called synthesis, is to take all the attributes over the original relation  $R$  and a minimal cover  $F$  for the FDs that hold over it and add a relation schema  $XA$  to the decomposition of  $R$  for each FD  $X \rightarrow A$  in  $F$ .

The resulting collection of relation schemas is in 3NF and preserves all FDs. If it is not a lossless-join decomposition of  $R$ , we can make it so by adding a relation schema that contains just those attributes that appear in some key. This algorithm gives us a lossless-join, dependency-preserving decomposition into 3NF and has polynomial complexity-----polynomial algorithms are available for computing minimal covers, and a key can be found in polynomial time (even though finding all keys is known to be NP-complete). The existence of a polynomial algorithm for obtaining a lossless-join, dependency-preserving decomposition into 3NF is surprising when we consider that testing whether a given schema is in 3NF is NP-complete.

As an example, consider a relation  $ABC$  with FDs  $F = \{A \rightarrow B, C \rightarrow B\}$ . The first step yields the relation schemas  $AB$  and  $BC$ . This is not a lossless-join decomposition of  $ABC$ ;  $AB \bowtie BC$  is  $B$ , and neither  $B \rightarrow A$  nor  $B \rightarrow C$  is in  $F^+$ . If we add a schema  $AC$ , we have the lossless-join property as well. Although the collection of relations  $AB, BC$ , and  $AC$  is a dependency-preserving, lossless-join decomposition of  $ABC$ , we obtained it through a process of *synthesis*, rather than through a process of repeated decomposition. We note that the decomposition produced by the synthesis approach heavily depends on the minimal cover used.

As another example of the synthesis approach, consider the Contracts relation with attributes  $CSJDPQV$  and the following FDs:

$$C \rightarrow CSJDPQV, SP \rightarrow C, SD \rightarrow P, \text{ and } J \rightarrow S.$$

This set of FDs is not a minimal cover, and so we must find one. We first replace  $G \rightarrow CSJDPQV$  with the FDs:



$$C \rightarrow S, C \rightarrow J, C \rightarrow IJ, C \twoheadrightarrow P, C \rightarrow Q, \text{ and } C \rightarrow V.$$

The FD  $C \rightarrow P$  is implied by  $C \rightarrow S$ ,  $C \twoheadrightarrow D$ , and  $SD \rightarrow P$ ; so we can delete it. The FD  $C \rightarrow S$  is implied by  $C \rightarrow J$  and  $J \twoheadrightarrow S$ ; so we can delete it. This leaves us with a minimal cover:

$$C \rightarrow J, C \rightarrow IJ, C \rightarrow Q, C \rightarrow V, JP \rightarrow C, SD \rightarrow P, \text{ and } J \twoheadrightarrow S.$$

Using the algorithm for ensuring dependency-preservation, we obtain the relational schema  $CJ, CD, CQ, CV, GJP, SDP$ , and  $JB$ . We can improve this schema by combining relations for which  $C$  is the key into  $CDJPQV$ . In addition, we have  $SDP$  and  $JS$  in our decomposition. Since one of these relations ( $CDJPQV$ ) is a superkey, we are done.

Comparing this decomposition with that obtained earlier in this section, we find they are quite close, with the only difference being that one of them has  $CDJPQV$  instead of  $CJP$  and  $CJDQV$ . In general, however, there could be significant differences.

## 19.7 SCHEMA REFINEMENT IN DATABASE DESIGN

We have seen how normalization can eliminate redundancy and discussed several approaches to normalizing a relation. We now consider how these ideas are applied in practice.

Database designers typically use a conceptual design methodology, such as ER design, to arrive at an initial database design. Given this, the approach of repeated decompositions to rectify instances of redundancy is likely to be the most natural use of PIs and normalization techniques.

In this section, we motivate the need for a schema refinement step following ER design. It is natural to ask whether we even need to decompose relations produced by translating an ER diagram. Should a good ER design not lead to a collection of relations free of redundancy problems? Unfortunately, ER design is a complex, subjective process, and certain constraints are not expressible in terms of ER diagrams. The examples in this section are intended to illustrate why decomposition of relations produced through ER design might be necessary.

### 19.7.1 Constraints on an Entity Set

Consider the Hourly\_Emps relation again. The constraint that attribute *ssn* is a key can be expressed as an FD:

$$\{ssn\} \rightarrow \{ssn, name, lot, rating, hourly\_wages, hours\_worked\}$$

For brevity, we write this FD as  $S \rightarrow SNLRWH$ , using a single letter to denote each attribute and omitting the set braces, but the reader should remember that both sides of an FD contain sets of attributes. In addition, the constraint that the *hourly\_wages* attribute is determined by the *rating* attribute is an FD:  $R \rightarrow W$ .

As we saw in Section 19.1.1, this FD led to redundant storage of rating wage associations. *It cannot be expressed in terms of the ER model. Only FDs that determine all attributes of a relation (i.e., key constraints) can be expressed in the ER model.* Therefore, we could not detect it when we considered Hourly\_Emps as an entity set during ER modeling.

We could argue that the problem with the original design was an artifact of a poor ER design, which could have been avoided by introducing an entity set called Wage\_Table (with attributes *rating* and *hourly\_wages*) and a relationship set Has\_Wages associating Hourly\_Emps and Wage\_Table. The point, however, is that we could easily arrive at the original design given the subjective nature of ER modeling. Having formal techniques to identify the problem with this design and guide us to a better design is very useful. The value of such techniques cannot be underestimated when designing large schemas—schemas with more than a hundred tables are not uncommon.

### 19.7.2 Constraints on a Relationship Set

The previous example illustrated how FDs can help to refine the subjective decisions made during ER design, but one could argue that the best possible ER diagram would have led to the same final set of relations. Our next example shows how functional information can lead to a set of relations unlikely to be arrived at solely through ER design.

We revisit an example from Chapter 2. Suppose that we have entity sets Parts, Suppliers, and Departments, as well as a relationship set Contracts that involves all of them. We refer to the schema for Contracts as *CQPSD*. A contract with contract id *C* specifies that a supplier *S* will supply some quantity *Q* of a part *P* to a department *J*. (We have added the contract id field *C* to the version of the Contracts relation discussed in Chapter 2.)

We might have a policy that a department purchases at most one part from any given supplier. Therefore, if there are several contracts between the same supplier and department, we know that the same part must be involved in all of them. This constraint is an FD,  $DS \rightarrow P$ .

Again we have redundancy and its associated problems. We can address this situation by decomposing Contracts into two relations with attributes *CQSD* and *3DP*. Intuitively, the relation *3DP* records the part supplied to a department by a supplier, and the relation *CQSD* records additional information about a contract. It is unlikely that we would arrive at such a design solely through ER modeling, since it is hard to formulate an entity or relationship that corresponds naturally to *CQSD*.

### 19.7.3 Identifying Attributes of Entities

This example illustrates how a careful examination of FDs can lead to a better understanding of the entities and relationships underlying the relational tables; in particular, it shows that attributes can easily be associated with the ‘wrong’ entity set during ER design. The ER diagram in Figure 19.11 shows a relationship set called *Works\_In* that is similar to the *Works\_In* relationship set of Chapter 2 but with an additional key constraint indicating that an employee can work in at most one department. (Observe the arrow connecting *Employees* to *Works\_In*.)

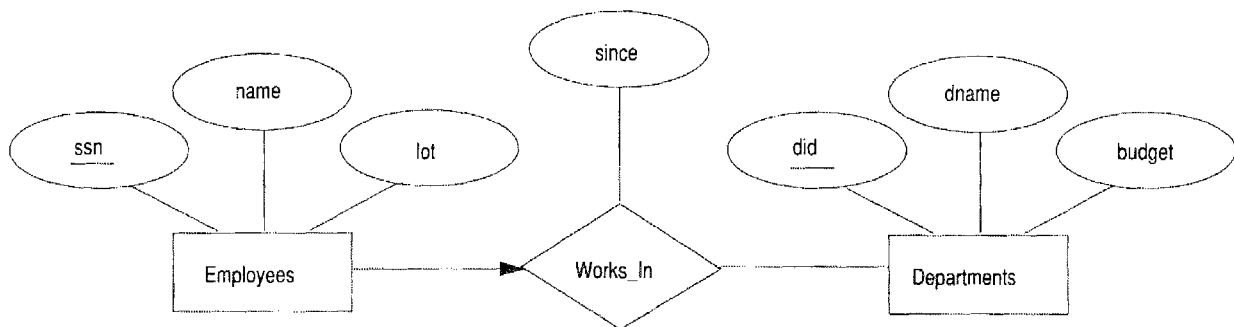


Figure 19.11. The *Works\_In* Relationship Set

Using the key constraint, we can translate this ER diagram into two relations:

*Workers*(*ssn*, *name*, *lot*, *d'id*, *since*)  
*Departments*(*did*, *dname*, *budget*)

The entity set *Employees* and the relationship set *Works\_In* are mapped to a single relation, *vWorkers*. This translation is based on the second approach discussed in Section 2.4.1.

Now suppose employees are assigned parking lots based on their department, and that all employees in a given department are assigned to the same lot. This constraint is not expressible with respect to the ER diagram of Figure 19.11. It is another example of an FD:  $did \rightarrow lot$ . The redundancy in this design can be eliminated by decomposing the Workers relation into two relations:

```
vWorkers2(ssn, name, did, since)
Dept_Lots(did, lot)
```

The new design has much to recommend it. We can change the lots associated with a department by updating a single tuple in the second relation (i.e., no update anomalies). We can associate a lot with a department even if it currently has no employees, without using *null* values (i.e., no deletion anomalies). We can add an employee to a department by inserting a tuple to the first relation even if there is no lot associated with the employee's department (i.e., no insertion anomalies).

Examining the two relations Departments and Dept\_Lots, which have the same key, we realize that a Departments tuple and a Dept\_Lots tuple with the same key value describe the same entity. This observation is reflected in the ER diagram shown in Figure 19.12.

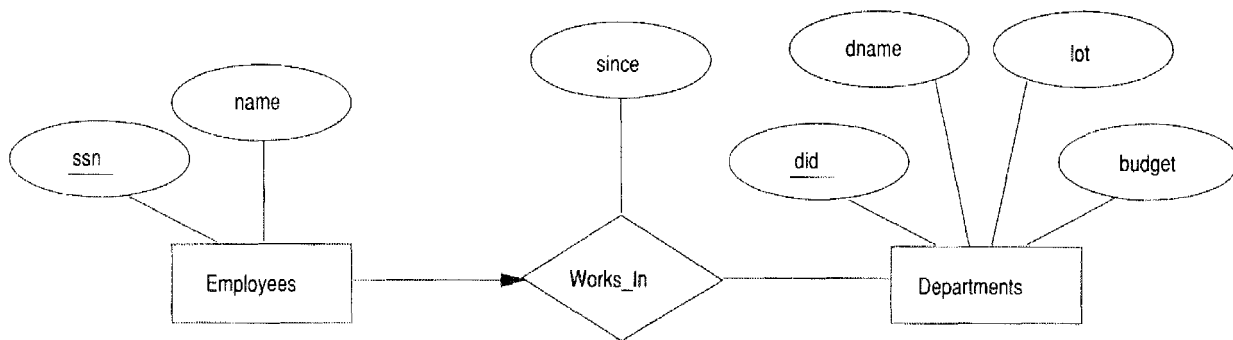


Figure 19.12 Refined Works\_In Relationship Set

Translating this diagram into the relational model would yield:

```
Workers2(ssn, name, did, since)
Departments(did, dname, budget, lot)
```

It seems intuitive to associate lots with employees; on the other hand, the lessons reveal that, in this example lots are really associated with departments. The subjective process of ER modeling could miss this point. The rigorous process of normalization would not.

### 19.7.4 Identifying Entity Sets

Consider a variant of the Reserves schema used in earlier chapters. Let Reserves contain attributes  $S$ ,  $B$ , and  $D$  as before, indicating that sailor  $S$  has a reservation for boat  $B$  on day  $D$ . In addition, let there be an attribute  $C$  denoting the credit card to which the reservation is charged. We use this example to illustrate how FD information can be used to refine an ER design. In particular, we discuss how FD information can help decide whether a concept should be modeled as an entity or as an attribute.

Suppose every sailor uses a unique credit card for reservations. This constraint is expressed by the FD  $S \rightarrow C$ . This constraint indicates that, in relation Reserves, we store the credit card number for a sailor as often as we have reservations for that sailor, and we have redundancy and potential update anomalies. A solution is to decompose Reserves into two relations with attributes  $SBD$  and  $SC$ . Intuitively, one holds information about reservations, and the other holds information about credit cards.

It is instructive to think about an ER design that would lead to these relations. One approach is to introduce an entity set called Credit\_Cards, with the sole attribute *cardno*, and a relationship set Has\_Card associating Sailors and Credit\_Cards. By noting that each credit card belongs to a single sailor, we can map Has\_Card and Credit\_Cards to a single relation with attributes  $SC$ . We would probably not model credit card numbers as entities if our main interest in card numbers is to indicate how a reservation is to be paid for; it suffices to use an attribute to model card numbers in this situation.

A second approach is to make *cardno* an attribute of Sailors. But this approach is not very natural—a sailor may have several cards, and we are not interested in all of them. Our interest is in the one card that is used to pay for reservations, which is best modeled as an attribute of the relationship Reserves.

A helpful way to think about the design problem in this example is that we first make *cardno* an attribute of Reserves and then refine the resulting tables by taking into account the FD information. (Whether we refine the design by adding *cardno* to the table obtained from Sailors or by creating a new table with attributes  $SC$  is a separate issue.)

## 19.8 OTHER KINDS OF DEPENDENCIES

FDs are probably the most common and important kind of constraint from the point of view of database design. However, there are several other kinds of dependencies. In particular, there is a well-developed theory for database

design using *multivalued dependencies* and *join dependencies*. By taking such dependencies into account, we can identify potential redundancy problems that cannot be detected using FDs alone.

This section illustrates the kinds of redundancy that can be detected using multivalued dependencies. Our main observation, however, is that simple guidelines (which can be checked using only FD reasoning) can tell us whether we even need to worry about complex constraints such as multivalued and join dependencies. We also comment on the role of *inclusion dependencies* in database design.

### 19.8.1 Multivalued Dependencies

Suppose that we have a relation with attributes *course*, *teacher*, and *book*, which we denote as *CTB*. The meaning of a tuple is that teacher *T* can teach course *C*, and book *B* is a recommended text for the course. There are no FDs; the key is *CTB*. However, the recommended texts for a course are independent of the instructor. The instance shown in Figure 19.13 illustrates this situation.

<i>course</i>	<i>teacher</i>	<i>book</i>
Physics101	Green	Mechanics
Physics101	Green	Optics
Physics101	Brown	Mechanics
Physics101	Brown	Optics
Math301	Green	Mechanics
Math301	Green	Vectors
Math301	Green	Geometry

Figure 19.13 BCNF Relation with Redundancy That Is Revealed by MVDs

Note three points here:

- The relation schema *CTB* is in BCNF; therefore we would not consider decomposing it further if we looked only at the FDs that hold over *(CTB)*.
- There is redundancy. The fact that Green can teach Physics101 is recorded once per recommended text for the course. Similarly, the fact that Optics is a text for Physics101 is recorded once per potential teacher.
- The redundancy can be eliminated by decomposing *CTB* into *CT* and *CE*.

The redundancy in this example is due to the constraint that the texts for a course are independent of the instructors, which cannot be expressed in terms

of FDs. This constraint is an example of a *multivalued dependency*, or MVD. Ideally, we should model this situation using two binary relationship sets, Instructors with attributes *CT* and Text with attributes *CB*. Because these are two essentially independent relationships, modeling them with a single ternary relationship set with attributes *CTE* is inappropriate. (See Section 2.5.3 for a further discussion of ternary versus binary relationships.) Given the subjectivity of ER design, however, we might create a ternary relationship. A careful analysis of the MVD information would then reveal the problem.

Let  $R$  be a relation schema and let  $X$  and  $Y$  be subsets of the attributes of  $R$ . Intuitively, the multivalued dependency  $X \twoheadrightarrow Y$  is said to hold over  $R$  if, in every legal instance  $r$  of  $R$ , each  $X$  value is associated with a set of  $Y$  values and this set is independent of the values in the other attributes.

Finally, if the MVD  $X \twoheadrightarrow Y$  holds over  $R$  and  $Z = R - XY$ , the following must be true for every legal instance  $r$  of  $R$ :

If  $t_1 \in r$ ,  $t_2 \in r$  and  $t_1.X = t_2.X$ , then there must be some  $t_3 \in r$  such that  $t_1.Y = t_3.Y$  and  $t_2.Z = t_3.Z$ ,

Figure 19.14 illustrates this definition. If we are given the first two tuples and told that the MVD  $X \twoheadrightarrow Y$  holds over this relation, we can infer that the relation instance must also contain the third tuple. Indeed, by interchanging the roles of the first two tuples—treating the first tuple as  $t_2$  and the second tuple as  $t_1$ —we can deduce that the tuple  $t_4$  must also be in the relation instance.

$X \mid Y \mid Z$			
$a$	$b_1$	$c_1$	— tuple $t_1$
$a$	$b_2$	$c_2$	— tuple $t_2$
$a$	$b_1$	$c_2$	— tuple $t_3$
$a$	$b_2$	$c_1$	— tuple $t_4$

Figure 19.14 Illustration of MVD Definition

This table suggests another way to think about MVDs: If  $X \twoheadrightarrow Y$  holds over  $R$ , then  $\pi_{YZ}(\sigma_{X=x}(R)) = \pi_Y(\sigma_{X=x}(R)) \times \pi_Z(\sigma_{X=x}(R))$  in every legal instance of  $R$ , for any value  $x$  that appears in the  $X$  column of  $R$ . In other words, consider groups of tuples in  $R$  with the same  $X$ -value. In each such group consider the projection onto the attributes  $YZ$ . This projection must be equal to the cross-product of the projections onto  $Y$  and  $Z$ . That is, for a given  $X$ -value, the  $Y$ -values and  $Z$ -values are independent. (From this definition it is easy to see that  $X \twoheadrightarrow Y$  will hold whenever  $X \twoheadrightarrow Z$  holds. If the FD  $X \rightarrow Y$

$Y$  holds, there is exactly one  $Y$ -value for a given  $X$ -value, and the conditions in the MVD definition hold trivially. The converse does not hold, as Figure 19.14 illustrates.)

Returning to our *CTB* example, the constraint that course texts are independent of instructors can be expressed as  $C \twoheadrightarrow T$ . In terms of the definition of MVDs, this constraint can be read as follows:

If (there is a tuple showing that)  $C$  is taught by teacher  $T$ ,  
 and (there is a tuple showing that)  $C$  has book  $B$  as text,  
 then (there is a tuple showing that)  $C$  is taught by  $T$  and has text  $B$ .

Given a set of FDs and MVDs, in general, we can infer that several additional FDs and MVDs hold. A sound and complete set of inference rules consists of the three Armstrong Axioms plus five additional rules. Three of the additional rules involve only MVDs:

- **MVD Complementation:** If  $X \twoheadrightarrow Y$ , then  $X \twoheadrightarrow R - XY$ .
- **MVD Augmentation:** If  $X \twoheadrightarrow Y$  and  $W \supseteq Z$ , then  $WX \twoheadrightarrow YZ$ .
- **MVD Transitivity:** If  $X \twoheadrightarrow Y$  and  $Y \twoheadrightarrow Z$ , then  $X \twoheadrightarrow (Z - Y)$ .

As an example of the use of these rules, since we have  $C \twoheadrightarrow T$  over *CTB*, MVD complementation allows us to infer that  $C \twoheadrightarrow CTB - CT$  as well, that is,  $C \twoheadrightarrow B$ . The remaining two rules relate FDs and MVDs:

- **Replication:** If  $X \rightarrow Y$ , then  $X \twoheadrightarrow Y$ .
- **Coalescence:** If  $X \twoheadrightarrow Y$  and there is a  $W$  such that  $W \cap Y$  is empty,  $W \rightarrow Z$ , and  $Y \supseteq Z$ , then  $X \rightarrow Z$ .

(Observe that replication states that every FD is also an MVD.)

## 19.8.2 Fourth Normal Form

Fourth Normal Form is a direct generalization of BCNF. Let  $R$  be a relation schema,  $X$  and  $Y$  be nonempty subsets of the attributes of  $R$ , and  $F$  be a set of dependencies that includes both FDs and MVDs.  $R$  is said to be in fourth normal form (4NF), if, for every nontrivial  $X \twoheadrightarrow Y$  that holds over  $R$ , one of the following statements is true:

- $Y \subseteq X$  or  $XY = R$ , or
- $X$  is a superkey.



In reading this definition, it is important to understand that the definition of a *key* has not changed.....the key must uniquely determine all attributes through FDs alone.  $X \twoheadrightarrow Y$  is a trivial MVD if  $Y \subseteq X \subseteq R$  or  $XY = R$ ; such MVDs always hold.

The relation  $CTB$  is not in 4NF because  $C \twoheadrightarrow T$  is a nontrivial MVD and  $C$  is not a key. We can eliminate the resulting redundancy by decomposing  $CTB$  into  $CT$  and  $CB$ ; each of these relations is then in 4NF.

To use MVD information fully, we must understand the theory of MVDs. However, the following result due to Date and Fagin identifies conditions—detected using only FD information!—under which we can safely ignore MVD information. That is, using MVD information in addition to the FD information will not reveal any redundancy. Therefore, if these conditions hold, we do not even need to identify all MVDs.

If a relation schema is in BCNF, and at least one of its keys consists of a single attribute, it is also in 4NF.

An important assumption is implicit in any application of the preceding result: *The set of FDs identified thus far is 'indeed the set of all FDs that hold over the relation.* This assumption is important because the result relies on the relation being in BCNF, which in turn depends on the set of FDs that hold over the relation.

We illustrate this point using an example. Consider a relation schema  $ABCD$  and suppose that the FD  $A \rightarrow BCD$  and the MVD  $B \twoheadrightarrow C$  are given. Considering only these dependencies, this relation schema appears to be a counterexample to the result. The relation has a simple key, appears to be in BCNF, and yet is not in 4NF because  $B \twoheadrightarrow C$  causes a violation of the 4NF conditions. Let us take a closer look.

$B$	$C$	$A$	$D$	
$b$	$c_1$	$a_1$	$d_1$	tuple $t_1$
$b$	$c_2$	$a_2$	$d_2$	tuple $t_2$
$b$	$c_1$	$a_2$	$d_2$	tuple $t_3$

Figure 19.15 Three Tuples from a Legal Instance of  $ABCD$

Figure 19.15 shows three tuples from an instance of  $ABCD$  that satisfies the given MVD  $B \twoheadrightarrow C$ . From the definition of an MVD, given tuples  $t_1$  and  $t_2$ , it follows that tuple  $t_3$  must also be included in the instance. Consider tuples  $t_2$  and  $t_3$ . From the given FD  $A \rightarrow BCD$  and the fact that these tuples have the

same  $A$ -value, we can deduce that  $c_1 = c_2$ . Therefore, we see that the FD  $B \rightarrow C$  must hold over  $ABCD$  whenever the FD  $A \rightarrow BCD$  and the MVD  $B \twoheadrightarrow C$  hold. If  $B \rightarrow C$  holds, the relation  $ABeD$  is not in **BCNF** (unless additional FDs make  $B$  a key)!

Thus, the apparent counterexample is really not a counterexample.....rather, it illustrates the importance of correctly identifying all FDs that hold over a relation. In this example,  $A \twoheadrightarrow BCD$  is not the only FD; the FD  $B \rightarrow C$  also holds but was not identified initially. Given a set of FDs and MVDs, the inference rules can be used to infer additional FDs (and MVDs); to apply the Date-Fagin result without first using the MVD inference rules, we must be certain that we have identified all the FDs.

In summary, the Date-Fagin result offers a convenient way to check that a relation is in 4NF (without reasoning about MVDs) if we are confident that we have identified all FDs. At this point, the reader is invited to go over the examples we have discussed in this chapter and see if there is a relation that is not in 4NF.

### 19.8.3 Join Dependencies

A join dependency is a further generalization of MVDs. A join dependency (JD)  $\bowtie \{R_1, \dots, R_n\}$  is said to hold over a relation  $R$  if  $R_1, \dots, R_n$  is a lossless-join decomposition of  $R$ .

An MVD  $X \twoheadrightarrow Y$  over a relation  $R$  can be expressed as the join dependency  $\bowtie \{XV, X(R, -Y)\}$ . As an example, in the *GTB* relation, the MVD  $C \twoheadrightarrow T$  can be expressed as the join dependency  $\bowtie \{Crr, CB\}$ .

Unlike FDs and MVDs, there is no set of sound and complete inference rules for JDs.

### 19.8.4 Fifth Normal Form

A relation schema  $R$  is said to be in fifth normal form (5NF) if, for every  $\Pi \bowtie \{R_1, \dots, R_n\}$  that holds over  $R$ , one of the following statements is true:

- $R_i = R$ , for some  $i$ , or
- The  $\Pi$  is implied by the set of those FDs over  $R$  in which the left side is a key for  $R$ .

The second condition deserves some explanation, since we have not presented inference rules for FDs and JDs taken together. Intuitively, we must be able to show that the decomposition of  $R$  into  $\{R_1, \dots, R_n\}$  is lossless-join whenever the key dependencies (FDs in which the left side is a key for  $R$ ) hold. (II)  $\bowtie \{R_1, \dots, R_n\}$  is a trivial JD if  $R_i = R$  for some  $i$ ; such a JD always holds.

The following result, also due to Date and Fagin, identifies conditions—again, detected using only FD information—under which we can safely ignore JD information:

If a relation schema is in 3NF and each of its keys consists of a single attribute, it is also in 5NF.

The conditions identified in this result are sufficient for a relation to be in 5NF but not necessary. The result can be very useful in practice because it allows us to conclude that a relation is in 5NF *without ever identifying the MVDs and JDs that may hold over the relation*.

## 19.8.5 Inclusion Dependencies

MVDs and JDs can be used to guide database design, as we have seen, although they are less common than FDs and harder to recognize and reason about. In contrast, inclusion dependencies are very intuitive and quite common. However, they typically have little influence on database design (beyond the ER design stage).

Informally, an inclusion dependency is a statement of the form that some columns of a relation are contained in other columns (usually of a second relation). A foreign key constraint is an example of an inclusion dependency; the referring column(s) in one relation must be contained in the primary key column(s) of the referenced relation. As another example, if  $R$  and  $S$  are two relations obtained by translating two entity sets that every  $R$  entity is also an  $S$  entity, we would have an inclusion dependency; projecting  $R$  on its key attributes yields a relation contained in the relation obtained by projecting  $S$  on its key attributes.

The main point to bear in mind is that we should not split groups of attributes that participate in an inclusion dependency. For example, if we have an inclusion dependency  $AB \subseteq CD$ , while decomposing the relation schema containing  $AB$ , we should ensure that at least one of the schemas obtained in the decomposition contains both  $A$  and  $B$ . Otherwise, we cannot check the inclusion dependency  $AB \subseteq CD$  without reconstructing the relation containing  $AB$ .

Most inclusion dependencies in practice are *key-based*, that is, involve only keys. Foreign key constraints are a good example of key-based inclusion dependencies. An ER diagram that involves ISA hierarchies (see Section 2.4.4) also leads to key-based inclusion dependencies. If all inclusion dependencies are key-based, we rarely have to worry about splitting attribute groups that participate in inclusion dependencies, since decompositions usually do not split the primary key. Note, however, that going from 3NF to BCNF always involves splitting some key (ideally not the primary key!), since the dependency guiding the split is of the form  $X \rightarrow A$  where  $A$  is part of a key.

## 19.9 CASE STUDY: THE INTERNET SHOP

Recall from Section 3.8 that DBDudes settled on the following schema:

Books(isbn: CHAR(10), title: CHAR(8), author: CHAR(80),  
 qty\_in\_stock: INTEGER, price: REAL, year\_published: INTEGER)  
 Customers(cid: INTEGER, name: CHAR(80), address: CHAR(200))  
 Orders(order\_num: INTEGER, isbn: CHAR(10), cid: INTEGER,  
 cardnum: CHAR(16), qty: INTEGER, order\_date: DATE, ship\_date: DATE)

DBDudes analyzes the set of relations for possible redundancy. The Books relation has only one key, (*isbn*), and no other functional dependencies hold over the table. Thus, Books is in BCNF. The Customers relation also has only one key, (*cid*), and no other functional dependencies hold over the table. Thus, Customers is also in BCNF.

DBDudes has already identified the pair  $\langle \text{order\_num}, \text{isbn} \rangle$  as the key for the Orders table. In addition, since each order is placed by one customer on one specific date with one specific credit card number, the following three functional dependencies hold:

$$\text{order\_num} \rightarrow \text{cid}, \text{order\_num} \rightarrow \text{order\_date}, \text{ and } \text{order\_num} \rightarrow \text{cardnum}$$

The experts at DBDudes conclude that Orders is not even in 3NF. (Can you see why?) They decide to decompose Orders into the following two relations:

Orders(order\_num, cid, order\_date, cardnum, and  
 Orderlists(order\_num, isbn, qty, ship\_date)

The resulting two relations, Orders and Orderlists, are both in BCNF, and the decomposition is lossless-join since *order\_num* is a key for (the new) Orders. The reader is invited to check that this decomposition is also dependency-preserving. For completeness, we give the SQL DDL for the Orders and Orderlists relations below:

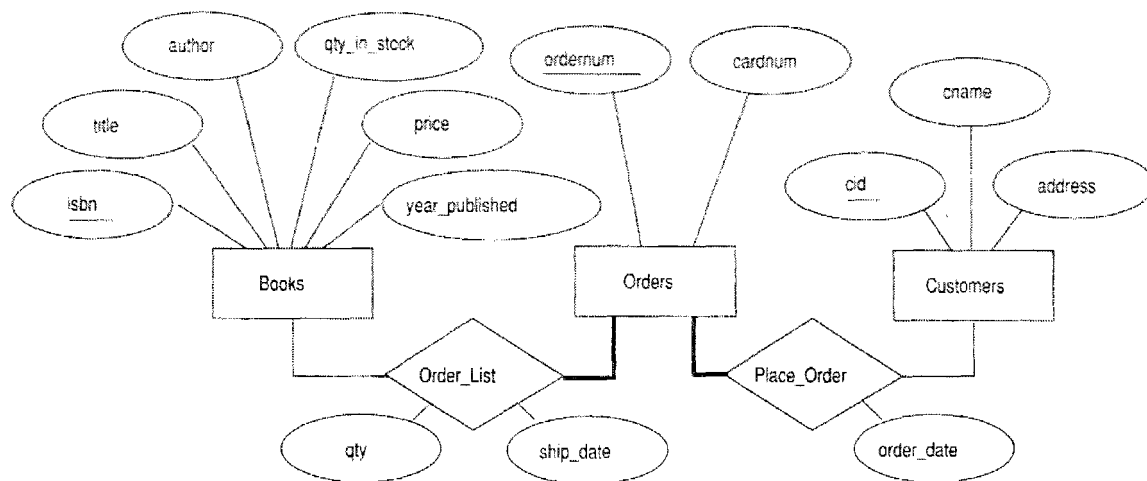


Figure 19.16 ER Diagram Reflecting the Final Design

```
CREATE TABLE Orders ( ordernum    INTEGER,
                      cid         INTEGER,
                      order_date  DATE,
                      cardnum     CHAR(16),
                      PRIMARY KEY (ordernum),
                      FOREIGN KEY (cid) REFERENCES Customers )
```

```
CREATE TABLE Orderlists ( ordernum    INTEGER,
                          isbn         CHAR(10),
                          qty          INTEGER,
                          ship_date    DATE,
                          PRIMARY KEY (ordernum, isbn),
                          FOREIGN KEY (isbn) REFERENCES Books)
```

Figure 19.16 shows an updated ER diagram that reflects the new design. Note that DBDudes could have arrived immediately at this diagram if they had made Orders an entity set instead of a relationship set right at the beginning. But at that time they did not understand the requirements completely, and it seemed natural to model Orders as a relationship set. This iterative refinement process is typical of real-life database design processes. As DBDudes has learned over time, it is rare to achieve an initial design that is not changed as a project progresses.

The DBDudes team celebrates the successful completion of logical database design and schema refinement by opening a bottle of champagne and charging it to B&N. After recovering from the celebration, they move on to the physical design phase.

## 19.10 REVIEW QUESTIONS

Answers to the review questions can be found in the listed sections.

- Illustrate redundancy and the problems that it can cause. Give examples of *insert*, *delete*, and *update* anomalies. Can *null* values help address these problems? Are they a complete solution? (Section 19.1.1)
- What is a *decomposition* and how does it address redundancy? What problems may be caused by the use of decompositions? (Sections 19.1.2 and 19.1.3)
- Define *functional dependencies*. How are *primary keys* related to FDs? (Section 19.2)
- When is an FD *implied* by a set  $F$  of FDs? Define *Armstrong's Axioms*, and explain the statement that "they are a sound and complete set of rules for FD inference." (Section 19.3)
- What is the *dependency closure*  $F^+$  of a set  $F$  of FDs? What is the *attribute closure*  $X^+$  of a set of attributes  $X$  with respect to a set of FDs  $F$ ? (Section 19.3)
- Define 1NF, 2NF, 3NF, and BCNF. What is the motivation for putting a relation in BCNF? What is the motivation for 3NF? (Section 19.4)
- When is the decomposition of a relation schema  $R$  into two relation schemas  $X$  and  $Y$  said to be a *lossless-join* decomposition? Why is this property so important? Give a necessary and sufficient condition to test whether a decomposition is lossless-join. (Section 19.5.1)
- When is a decomposition said to be *dependency-preserving*? Why is this property useful? (Section 19.5.2)
- Describe how we can obtain a lossless-join decomposition of a relation into BCNF. Give an example to show that there may not be a dependency-preserving decomposition into BCNF. Illustrate how a given relation could be decomposed in different ways to arrive at several alternative decompositions, and discuss the implications for database design. (Section 19.6.1)
- Give an example that illustrates how a collection of relations in BCNF could have redundancy even though each relation, by itself, is free from redundancy. (Section 19.6.1)
- What is a *minimal cover* for a set of FDs? Describe an algorithm for computing the minimal cover of a set of FDs, and illustrate it with an example. (Section 19.6.2)

- Describe how the algorithm for lossless-join decomposition into BCNF can be adapted to obtain a lossless-join, dependency-preserving decomposition into 3NF. Describe the alternative *synthesis* approach to obtaining such a decomposition into 3NF. Illustrate both approaches using an example. (Section 19.6.2)
- Discuss how schema refinement through dependency analysis and normalization can improve schemas obtained through ER design. (Section 19.7)
- Define *multivalued dependencies*, *join dependencies*, and *inclusion dependencies*. Discuss the use of such dependencies for database design. Define 4NF and 5NF, and explain how they prevent certain kinds of redundancy that BCNF does not eliminate. Describe tests for 4NF and 5NF that use only FDs. What key assumption is involved in these tests? (Section 19.8)

## EXERCISES

Exercise 19.1 Briefly answer the following questions:

1. Define the term *functional dependency*.
2. Why are some functional dependencies called *trivial*?
3. Give a set of FDs for the relation schema  $R(A, B, C, D)$  with primary key  $AB$  under which  $R$  is in 1NF but not in 2NF.
4. Give a set of FDs for the relation schema  $R(A, B, C, D)$  with primary key  $AB$  under which  $R$  is in 2NF but not in 3NF.
5. Consider the relation schema  $R(A, B, C)$ , which has the FD  $B \rightarrow C$ . If  $A$  is a candidate key for  $R$ , is it possible for  $R$  to be in BCNF? If so, under what conditions? If not, explain why not.
6. Suppose we have a relation schema  $R(A, B, C)$  representing a relationship between two entity sets with keys  $A$  and  $B$ , respectively, and suppose that  $R$  has (among others) the FDs  $A \twoheadrightarrow B$  and  $B \rightarrow A$ . Explain what such a pair of dependencies means (i.e., what they imply about the relationship that the relation models).

Exercise 19.2 Consider a relation  $R$  with five attributes  $ABCDE$ . You are given the following dependencies:  $A \rightarrow B$ ,  $BE \rightarrow E$ , and  $ED \rightarrow A$ .

1. List all keys for  $R$ .
2. Is  $R$  in 3NF?
3. Is  $R$  in BCNF?

Exercise 19.3 Consider the relation shown in Figure 19.17.

1. List all the functional dependencies that this relation instance satisfies.
2. Assume that the value of attribute  $Z$  of the last record in the relation is changed from  $z_3$  to  $z_2$ . Now list all the functional dependencies that this relation instance satisfies.

Exercise 19.4 Assume that you are given a relation with attributes  $ABCD$ .

$X$	$Y$	$Z$
$x_1$	$y_1$	$z_1$
$x_1$	$y_1$	$z_2$
$x_2$	$y_1$	$z_1$
$x_2$	$y_1$	$z_3$

Figure 19.17 Relation for Exercise 19.3.

1. Assume that no record has NULL values. Write an SQL query that checks whether the functional dependency  $A \rightarrow B$  holds.
2. Assume again that no record has NULL values. Write an SQL assertion that enforces the functional dependency  $A \rightarrow B$ .
3. Let us now assume that records could have NULL values. Repeat the previous two questions under this assumption.

**Exercise 19.5** Consider the following collection of relations and dependencies. Assume that each relation is obtained through decomposition from a relation with attributes  $ABCDEFGHI$  and that all the known dependencies over relation  $ABCDEFGHI$  are listed for each question. (The questions are independent of each other, obviously, since the given dependencies over  $ABCDEFGHI$  are different.) For each (sub)relation: (a) State the strongest normal form that the relation is in. (b) If it is not in BCNF, decompose it into a collection of BCNF relations.

1.  $R_1(A, C, B, D, E)$ ,  $A \rightarrow B$ ,  $C \rightarrow D$
2.  $R_2(A, B, F)$ ,  $AC \rightarrow B$ ,  $B \rightarrow F$
3.  $R_3(A, D)$ ,  $D \rightarrow G$ ,  $G \rightarrow H$
4.  $R_4(D, C, H, G)$ ,  $A \rightarrow I$ ,  $I \rightarrow A$
5.  $R_5(A, I, C, B)$

**Exercise 19.6** Suppose that we have the following three tuples in a legal instance of a relation schema  $S$  with three attributes  $ABC$  (listed in order):  $(1,2,3)$ ,  $(4,2,3)$ , and  $(5,3,3)$ .

1. Which of the following dependencies can you infer does *not* hold over schema  $S$ ?  
(a)  $A \rightarrow B$ , (b)  $BC \rightarrow A$ , (c)  $B \rightarrow C$
2. Can you identify any dependencies that hold over  $S$ ?

**Exercise 19.7** Suppose you are given a relation  $R$  with four attributes  $ABCD$ . For each of the following sets of FDs, assuming those are the only dependencies that hold for  $R$ , do the following: (a) Identify the candidate key(s) for  $R$ . (b) Identify the best normal form that  $R$  satisfies (1NF, 2NF, 3NF, or BCNF). (c) If  $R$  is not in BCNF, decompose it into a set of BCNF relations that preserve the dependencies.

1.  $C \rightarrow D$ ,  $C \rightarrow A$ ,  $B \rightarrow C$
2.  $B \rightarrow C$ ,  $D \rightarrow A$
3.  $ABC \rightarrow D$ ,  $D \rightarrow A$
4.  $A \rightarrow B$ ,  $BC \rightarrow D$ ,  $A \rightarrow C$
5.  $AB \rightarrow C$ ,  $AB \rightarrow D$ ,  $C \rightarrow A$ ,  $D \rightarrow B$



**Exercise 19.8** Consider the attribute set  $R = ABCDEGH$  and the FD set  $F = \{AB \rightarrow C, AC \rightarrow B, AD \rightarrow E, B \rightarrow D, Be \rightarrow A, B \rightarrow G\}$ .

- For each of the following attribute sets, do the following: (i) Compute the set of dependencies that hold over the set and write down a minimal cover. (ii) Name the strongest normal form that is not violated by the relation containing these attributes. (iii) Decompose it into a collection of BCNF relations if it is not in BCNF.

(a)  $ABC$ , (b)  $ABCD$ , (c)  $ABCEG$ , (d)  $DC:BGH$ , (e)  $ACEH$

- Which of the following decompositions of  $R = ABCDEG$ , with the same set of dependencies  $F$ , is (a) dependency-preserving? (b) lossless-join?

(a)  $\{AB, BC, ABDE, EG\}$

(b)  $\{ABC, ACDE, ADG\}$

**Exercise 19.9** Let  $R$  be decomposed into  $R_1, R_2, \dots, R_n$ . Let  $F$  be a set of FDs on  $R$ .

- Define what it means for  $F$  to be preserved in the set of decomposed relations.
- Describe a polynomial-time algorithm to test dependency-preservation.
- Projecting the FDs stated over a set of attributes  $X$  onto a subset of attributes  $Y$  requires that we consider the closure of the FDs. Give an example where considering the closure is important in testing dependency-preservation, that is, considering just the given FDs gives incorrect results.

**Exercise 19.10** Suppose you are given a relation  $R(A,B,C,D)$ . For each of the following sets of FDs, assuming they are the only dependencies that hold for  $R$ , do the following: (a) Identify the candidate key(s) for  $R$ . (b) State whether or not the proposed decomposition of  $R$  into smaller relations is a good decomposition and briefly explain why or why not.

- $B \rightarrow C, D \rightarrow A$ ; decompose into  $BC$  and  $AD$ .
- $AB \rightarrow C, C \rightarrow A, C \rightarrow D$ ; decompose into  $ACD$  and  $Be$ .
- $A \rightarrow BC, C \rightarrow AD$ ; decompose into  $ABC$  and  $AD$ .
- $A \rightarrow B, B \rightarrow C, C \rightarrow D$ ; decompose into  $AB$  and  $ACD$ .
- $A \rightarrow B, B \rightarrow C, C \rightarrow D$ ; decompose into  $AB, AD$  and  $CD$ .

**Exercise 19.11** Consider a relation  $R$  that has three attributes  $ABC$ . It is decomposed into relations  $R_1$  with attributes  $AB$  and  $R_2$  with attributes  $Be$ .

- State the definition of a lossless-join decomposition with respect to this example. Answer this question concisely by writing a relational algebra equation involving  $R, R_1$ , and  $R_2$ .
- Suppose that  $B \twoheadrightarrow C$ . Is the decomposition of  $R$  into  $R_1$  and  $R_2$  lossless-join? Reconcile your answer with the observation that neither of the FDs  $H_1 \rightarrow R_1$  nor  $R_1 \rightarrow R_2$  hold, in light of the simple test offering a necessary and sufficient condition for lossless-join decomposition into two relations in Section 15.6.1.
- If you are given the following instances of  $R_1$  and  $R_2$ , what can you say about the instance of  $R$  from which these were obtained? Answer this question by listing tuples that are definitely in  $R$  and tuples that are possibly in  $R$ .

Instance of  $R_1 = \{(5,1), (6,1)\}$

Instance of  $R_2 = \{(1,8), (1,9)\}$

Can you say that attribute  $B$  definitely is or is not a key for  $R$ ?

**Exercise 19.12** Suppose that we have the following four tuples in a relation  $S$  with three attributes  $ABC$ :  $(1,2,3)$ ,  $(4,2,3)$ ,  $(5,3,3)$ ,  $(5,3,4)$ . Which of the following functional ( $\rightarrow$ ) and multivalued ( $\twoheadrightarrow$ ) dependencies can you infer does *not* hold over relation  $S$ ?

1.  $A \rightarrow B$
2.  $A \twoheadrightarrow B$
3.  $BC \rightarrow A$
4.  $BC \twoheadrightarrow A$
5.  $B \rightarrow C$
6.  $B \twoheadrightarrow C$

**Exercise 19.13** Consider a relation  $R$  with five attributes  $ABCDE$ .

1. For each of the following instances of  $R$ , state whether it violates (a) the FD  $BE \rightarrow D$  and (b) the MVD  $BE \twoheadrightarrow D$ :
  - (a)  $\{ \}$  (i.e., empty relation)
  - (b)  $\{(0,2,3,4,5), (2,0,3,5,5)\}$
  - (c)  $\{(0,2,3,4,5), (2,0,3,5,5), (0,2,3,4,6)\}$
  - (d)  $\{(0,2,3,4,5), (2,0,3,4,5), (0,2,3,6,5)\}$
  - (e)  $\{(0,2,3,4,5), (2,0,3,7,5), (0,2,3,4,6)\}$
  - (f)  $\{(0,2,3,4,5), (2,0,3,4,5), (0,2,3,6,5), (0,2,3,6,6)\}$
  - (g)  $\{(0,2,3,4,5), (0,2,3,6,5), (0,2,3,6,6), (0,2,3,4,6)\}$
2. If each instance for  $R$  listed above is legal, what can you say about the FD  $A \rightarrow B$ ?

**Exercise 19.14** JDs are motivated by the fact that sometimes a relation that cannot be decomposed into two smaller relations in a lossless-join manner can be so decomposed into three or more relations. An example is a relation with attributes *supplier*, *part*, and *project*, denoted  $SPJ$ , with no FDs or MVDs. The JD  $\{SP, PJ, JS\}$  holds.

From the JD, the set of relation schemes  $SP$ ,  $PJ$ , and  $JS$  is a lossless-join decomposition of  $SPJ$ . Construct an instance of  $SPJ$  to illustrate that no two of these schemes suffice.

**Exercise 19.15** Answer the following questions

1. Prove that the algorithm shown in Figure 19.4 correctly computes the attribute closure of the input attribute set  $X$ .
2. Describe a linear-time (in the size of the set of FDs, where the size of each FD is the number of attributes involved) algorithm for finding the attribute closure of a set of attributes with respect to a set of FDs. Prove that your algorithm correctly computes the attribute closure of the input attribute set.

**Exercise 19.16** Let us say that an FD  $X \rightarrow Y$  is *simple* if  $Y$  is a single attribute.

1. Replace the FD  $AB \rightarrow CD$  by the smallest equivalent collection of simple FDs.
2. Prove that every FD  $X \rightarrow Y$  in a set of FDs  $F$  can be replaced by a set of simple FDs such that  $F^+$  is equal to the closure of the new set of FDs.

**Exercise 19.17** Prove that Armstrong's Axioms are sound and complete for FD inference. That is, show that repeated application of these axioms on a set  $F$  of FDs produces exactly the dependencies in  $P^+$ .

**Exercise 19.18** Consider a relation  $R$  with attributes  $ABCDE$ . Let the following FDs be given:  $A \rightarrow BC$ ,  $BE \rightarrow E$ , and  $E \rightarrow DA$ . Similarly, let  $S$  be a relation with attributes  $ABCDE$  and let the following FDs be given:  $A \rightarrow BC$ ,  $B \rightarrow E$ , and  $E \rightarrow DA$ . (Only the second dependency differs from those that hold over  $R$ .) You do not know whether or which other (join) dependencies hold.

1. Is  $R$  in BCNF?
2. Is  $R$  in 4NF?
3. Is  $R$  in 5NF?
4. Is  $S$  in BCNF?
5. Is  $S$  in 4NF?
6. Is  $S$  in 5NF?

**Exercise 19.19** Let  $R$  be a relation schema with a set  $F$  of FDs. Prove that the decomposition of  $R$  into  $H_1$  and  $R_2$  is lossless-join if and only if  $F^+$  contains  $H_1 \cap R_2 \rightarrow R_1$  or  $R_1 \cap R_2 \rightarrow R_2$ .

**Exercise 19.20** Consider a schema  $R$  with FDs  $F$  that is decomposed into schemas with attributes  $X$  and  $Y$ . Show that this is dependency-preserving if  $F' \subseteq (F_X \cup F_Y)^+$ .

**Exercise 19.21** Prove that the optimization of the algorithm for lossless-join, dependency-preserving decomposition into 3NF relations (Section 19.6.2) is correct.

**Exercise 19.22** Prove that the 3NF synthesis algorithm produces a lossless-join decomposition of the relation containing all the original attributes.

**Exercise 19.23** Prove that an MVD  $X \twoheadrightarrow Y$  over a relation  $R$  can be expressed as the join dependency  $\bowtie \{XY, X(R - Y)\}$ .

**Exercise 19.24** Prove that, if  $R$  has only one key, it is in BCNF if and only if it is in 3NF.

**Exercise 19.25** Prove that, if  $R$  is in 3NF and every key is simple, then  $R$  is in 4NF.

**Exercise 19.26** Prove these statements:

1. If a relation schema is in BCNF and at least one of its keys consists of a single attribute, it is also in 4NF.
2. If a relation schema is in 3NF and each key has a single attribute, it is also in 5NF.

**Exercise 19.27** Give an algorithm for testing whether a relation schema is in BCNF. The algorithm should be polynomial in the size of the set of given FDs. (The 'size' is the sum over all FDs of the number of attributes that appear in the FD.) Is there a polynomial algorithm for testing whether a relation schema is in 3NF?

**Exercise 19.28** Give an algorithm for testing whether a relation schema is in BCNF. The algorithm should be polynomial in the size of the set of given FDs. (The 'size' is the sum over all FDs of the number of attributes that appear in the FD.) Is there a polynomial algorithm for testing whether a relation schema is in 3NF?

**Exercise 19.29** Prove that the algorithm for decomposing a relation schema with a set of FDs into a collection of BCNF relation schemas as described in Section 19.6.1 is correct (i.e., it produces a collection of BCNF relations, and is lossless-join) and terminates.

## BIBLIOGRAPHIC NOTES

Textbook presentations of dependency theory and its use in database design include [3, 45, 501, 509, 747]. Good survey articles on the topic include [755, 415].

FDs were introduced in [187], along with the concept of 3NF, and axioms for inferring FDs were presented in [38]. BCNF was introduced in [188]. The concept of a legal relation instance and dependency satisfaction are studied formally in [328]. FDs were generalized to semantic data models in [768].

Finding a key is shown to be NP-complete in [497]. Lossless-join decompositions were studied in [28, 502, 627]. Dependency-preserving decompositions were studied in [74]. [81] introduced minimal covers. Decomposition into 3NF is studied by [81, 98] and decomposition into BCNF is addressed in [742]. [412] shows that testing whether a relation is in 3NF is NP-complete. [253] introduced 4NF and discussed decomposition into 4NF. Fagin introduced other normal forms in [254] (project-join normal form) and [255] (domain-key normal form). In contrast to the extensive study of vertical decompositions, there has been relatively little formal investigation of horizontal decompositions. [209] investigates horizontal decomposition.

IVDs were discovered independently by Delobel [211], Fagin [253], and Zaniolo [789]. Axioms for FDs and MVDs were presented in [73]. [593] shows that there is no axiomatization for JDs, although [662] provides an axiomatization for a more general class of dependencies. The sufficient conditions for 4NF and 5NF in terms of FDs that were discussed in Section 19.8 are from [205]. An approach to database design that uses dependency information to construct sample relation instances is described in [508, 509].