

## Constraints: Functional Dependencies

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When we get a relational schema,

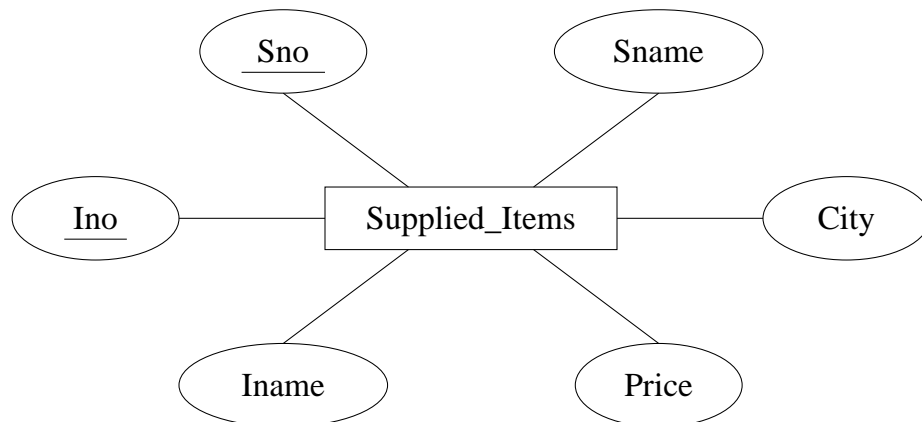
⇒ **how do we know if its any good?**

⇒ **what to watch for?**

- what are the allowed instances of the schema?
- does the structure capture the data?
  - ⇒ too hard to query?
  - ⇒ too hard to **update**?
  - ⇒ redundant information all over the place?

## Change Anomalies

Assume we are given the E-R diagram



## Change Anomalies (cont.)

Supplied\_Items

<u>Sno</u>	Sname	City	<u>Ino</u>	Iname	Price
S1	Magna	Ajax	I1	Bolt	0.50
S1	Magna	Ajax	I2	Nut	0.25
S1	Magna	Ajax	I3	Screw	0.30
S2	Budd	Hull	I3	Screw	0.40

Problems:

- 1 Update problems (Changing name of supplier)
- 2 Insert problems (New item w/o supplier)
- 3 Delete problems (Budd no longer supplies screws)
- 4 Likely increase in space requirements

## Change Anomalies (cont.)

Compare to

Sno	Sname	City
S1	Magna	Ajax
S2	Budd	Hull

Ino	Iname
I1	Bolt
I2	Nut
I3	Screw

Sno	Ino	Price
S1	I1	0.50
S1	I2	0.25
S1	I3	0.30
S2	I3	0.40

Decomposition seems to be better...

## Change Anomalies (cont.)

But other extreme is also undesirable

⇒ information about relationships can be lost

Sno
S1
S2

Sname
Magna
Budd

City
Ajax
Hull

Inum
I1
I2
I3

Iname
Bolt
Nut
Screw

Price
0.50
0.25
0.30
0.40

... so how do we know how much can we decompose?

## How to Find and Fix Anomalies?

**Detection:** How do we know an *anomaly* exists?

(certain families) of **Integrity Constraints** postulate regularities in schema instances that lead to anomalies.

**Repair** How can we *fix* it?

Certain **Schema Decompositions** avoid the anomalies while retaining *all information* in the instances.

## Integrity Constraints

**Idea:** allow only **well-behaved** instances of the schema

⇒ the relational structure (= selection of relations)

is often not sufficient to capture all of these.

- restrict values of an attribute
- describe dependencies between attributes
  - ⇒ in a single relation (bad)
  - ⇒ between relations (good)
- postulate the existence of values in the database
- ...

Dependencies between attributes in a single relation lead to improvements in schema design.

## Functional Dependencies (FDs)

**Idea:** to express the fact that in a relation **schema**

(values of) a set of attributes uniquely **determine**  
(values of) another set of attributes.

**Definition:** Let  $R$  be a relation schema, and  $X, Y \subseteq R$  sets of attributes. The **functional dependency**  $X \rightarrow Y$  is the formula

$$\forall v_1, \dots, v_k, w_1, \dots, w_k. R(v_1, \dots, v_k) \wedge R(w_1, \dots, w_k) \wedge \left( \bigwedge_{j \in X} v_j = w_j \right) \rightarrow \left( \bigwedge_{i \in Y} v_i = w_i \right)$$

We say that (the set of attributes)  $X$  **functionally determines**  $Y$  (in  $R$ ).

## Implication for FDs

How do we know what additional FDs hold in a schema?

A set  $F$  *logically implies* a FD  $X \rightarrow Y$  if  $X \rightarrow Y$  holds  
in *all instances* of  $R$  that satisfy  $F$ .

The **closure** of  $F^+$  of  $F$  is the set of all functional dependencies that  
are *logically implied* by  $F$

**Clearly:**  $F \subseteq F^+$ , but what else is in  $F^+$ ?

**For Example:**

$$F = \{A \rightarrow B, B \rightarrow C\} \text{ then } F^+ = \{A \rightarrow B, B \rightarrow C, A \rightarrow C\}$$

## Examples of Functional Dependencies

Consider the following relation schema:

EmpProj						
SIN	PNum	Hours	EName	PName	PLoc	Allowance

- SIN determines employee name  
 $SIN \rightarrow EName$
- project number determines project name and location  
 $PNum \rightarrow PName, PLoc$
- allowances are always the same for the same number of hours at the same location  
 $PLoc, Hours \rightarrow Allowance$

## Reasoning About FDs

Logical implications can be derived by using inference rules called **Armstrong's axioms**

- (reflexivity)  $Y \subseteq X \Rightarrow X \rightarrow Y$
- (augmentation)  $X \rightarrow Y \Rightarrow XZ \rightarrow YZ$
- (transitivity)  $X \rightarrow Y, Y \rightarrow Z \Rightarrow X \rightarrow Z$

The axioms are

- sound (anything derived from  $F$  is in  $F^+$ )
- complete (anything in  $F^+$  can be derived)

Additional rules can be derived

- (union)  $X \rightarrow Y, X \rightarrow Z \Rightarrow X \rightarrow YZ$
- (decomposition)  $X \rightarrow YZ \Rightarrow X \rightarrow Y$

## Reasoning (example)

**Example:**  $F = \{$   
   $SIN, PNum \rightarrow Hours$   
   $SIN \rightarrow EName$   
   $PNum \rightarrow PName, PLoc$   
   $PLoc, Hours \rightarrow Allowance \}$

A derivation of  $SIN, PNum \rightarrow Allowance$ :

1.  $SIN, PNum \rightarrow Hours$  ( $\in F$ )
2.  $PNum \rightarrow PName, PLoc$  ( $\in F$ )
3.  $PLoc, Hours \rightarrow Allowance$  ( $\in F$ )
4.  $SIN, PNum \rightarrow PNum$  (reflexivity)
5.  $SIN, PNum \rightarrow PName, PLoc$  (transitivity, 4 and 2)
6.  $SIN, PNum \rightarrow PLoc$  (decomposition, 5)
7.  $SIN, PNum \rightarrow PLoc, Hours$  (union, 6, 1)
8.  $SIN, PNum \rightarrow Allowance$  (transitivity, 7 and 3)

## Keys: formal definition

**Definition:**

- $K \subseteq R$  is a **superkey** for relation schema  $R$  if dependency  $K \rightarrow R$  holds on  $R$ .
- $K \subseteq R$  is a **candidate key** for relation schema  $R$  if  $K$  is a superkey and no subset of  $K$  is a superkey.

**Primary Key** = a candidate key chosen by the DBA.

## Efficient Reasoning

How to figure out if an FD is implied by  $F$  **quickly**?

$\Rightarrow$  a mechanical and more efficient way of using Armstrong's axioms:

```
function ComputeX+(X, F)
begin
  X+ := X;
  while true do
    if there exists (Y → Z) ∈ F such that
      (1) Y ⊆ X+, and
      (2) Z ⊄ X+
    then X+ := X+ ∪ Z
    else exit;
  return X+;
end
```

## Efficient Reasoning (cont.)

Let  $R$  be a relational schema and  $F$  a set of functional dependencies on  $R$ . Then

**Theorem:**  $X$  is a superkey of  $R$  if and only if

$$ComputeX^+(X, F) = R$$

**Theorem:**  $X \rightarrow Y \in F^+$  if and only if

$$Y \subseteq ComputeX^+(X, F)$$

## Computing a Decomposition

### Decomposition

Let  $R$  be a relation schema (= set of attributes). The collection  $\{R_1, \dots, R_n\}$  of relation schemas is a **decomposition** of  $R$  if

$$R = R_1 \cup R_2 \cup \dots \cup R_n$$

A good decomposition does not

- lose information
- complicate checking of constraints
- contain anomalies (or at least contains fewer anomalies)

## Lossless-Join Decompositions (cont.)

But computing the natural join of SGM and AM produces

Student	Assignment	Group	Mark
Ann	A1	G1	80
Ann	A2	G3	60
Ann	A1	G3	60 !
Bob	A2	G2	60 !
Bob	A1	G2	60

... and we get extra data (**spurious tuples**) and would therefore lose information if we were to replace Marks by SGM and AM.

If re-joining SGM and AM would **always** produce exactly the tuples in Marks, then we call SGM and AM a **lossless-join decomposition**.

## Lossless-Join Decompositions

We should be able to construct the instance of the original table from the instances of the tables in the decomposition

**Example:** Consider replacing

Marks

Student	Assignment	Group	Mark
Ann	A1	G1	80
Ann	A2	G3	60
Bob	A1	G2	60

by decomposing to two tables

SGM			AM	
Student	Group	Mark	Assignment	Mark
Ann	G1	80	A1	80
Ann	G3	60	A2	60
Bob	G2	60	A1	60

## Lossless-Join Decompositions (cont.)

A decomposition  $\{R_1, R_2\}$  of  $R$  is lossless if and only if the common attributes of  $R_1$  and  $R_2$  form a superkey for either schema, that is

$$R_1 \cap R_2 \rightarrow R_1 \text{ or } R_1 \cap R_2 \rightarrow R_2$$

**Example:** In the previous example we had

$$R = \{Student, Assignment, Group, Mark\},$$

$$F = \{(Student, Assignment \rightarrow Group, Mark)\},$$

$$R_1 = \{Student, Group, Mark\},$$

$$R_2 = \{Assignment, Mark\}$$

$\Rightarrow$  decomposition  $\{R_1, R_2\}$  is lossy because

$$R_1 \cap R_2 (= \{M\}) \text{ is not a superkey of either SGM or AM}$$

## Dependency Preservation

How do we test/enforce constraints on the decomposed schema?

**Example:** A table for a company database could be

R	FD1: Proj $\rightarrow$ Dept,
Proj   Dept   Div	FD2: Dept $\rightarrow$ Div, and
	FD3: Proj $\rightarrow$ Div

and two decompositions

$D_1 = \{R1[Proj, Dept], R2[Dept, Div]\}$

$D_2 = \{R1[Proj, Dept], R3[Proj, Div]\}$

Both are lossless. (Why?)

## Avoiding Anomalies

What is a “good” relational database schema?

Rule of thumb: Independent facts in separate tables:

“Each relation schema should consist of a **primary key** and a **set of mutually independent attributes**”

$\Rightarrow$  achieved by transformation of a schema to a **normal form**

**Goals:**

- Intuitive and straightforward changes
- Anomaly-free/Nonredundant representation of data

**We discuss:**

- Boyce-Codd Normal Form (BCNF)
- Third Normal Form (3NF)

... both based on the notion of **functional dependency**

## Dependency Preservation (cont.)

Which decomposition is *better*?

- Decomposition  $D_1$  lets us test FD1 on table R1 and FD2 on table R2; if they are both satisfied, FD3 is automatically satisfied.
- In decomposition  $D_2$  we can test FD1 on table R1 and FD3 on table R3. Dependency FD2 is an **interrelational constraint**: testing it requires joining tables R1 and R3.

$\Rightarrow D_1$  is better!

A decomposition  $D = \{R_1, \dots, R_n\}$  of  $R$  is **dependency preserving** if there is an equivalent set  $F'$  of functional dependencies, none of which is interrelational in  $D$ .

## Boyce-Codd Normal Form (BCNF)

Schema  $R$  is in **BCNF** (w.r.t.  $F$ ) if and only if whenever  $(X \rightarrow Y) \in F^+$  and  $XY \subseteq R$ , then either

- $(X \rightarrow Y)$  is trivial (i.e.,  $Y \subseteq X$ ), or
- $X$  is a superkey of  $R$

A database schema  $\{R_1, \dots, R_n\}$  is in BCNF if each relation schema  $R_i$  is in BCNF.

Formalization of the goal that **independent relationships** are stored in **separate tables**.

## BCNF (cont.)

Why does BCNF avoid redundancy?

For the schema *Supplied\_Items* we had a FD:

$$\text{Sno} \rightarrow \text{Sname, City}$$

Therefore: supplier name “Magna” and city “Ajax” must be repeated for each item supplied by supplier S1.

Assume the above FD holds over a schema *R* that is in BCNF. Then:

- Sno is a superkey for *R*
- each Sno value appears on one row only
- no need to repeat Sname and City values

## Lossless-Join BCNF Decomposition

```
function ComputeBCNF(R, F)  
begin  
  Result := {R};  
  while some Ri ∈ Result and (X → Y) ∈ F+  
    violate the BCNF condition do begin  
    Replace Ri by Ri − (Y − X);  
    Add {X, Y} to Result;  
  end;  
  return Result;  
end
```

## Lossless-Join BCNF Decomposition

- No *efficient* procedure to do this exists.
- Results depend on sequence of FDs used to decompose the relations.
- It is possible that no lossless join dependency preserving BCNF decomposition exists:

Consider  $R = \{A, B, C\}$  and  $F = \{AB \rightarrow C, C \rightarrow B\}$ .

## Third Normal Form (3NF)

Schema *R* is in **3NF** (w.r.t. *F*) if and only if whenever  $(X \rightarrow Y) \in F^+$  and  $XY \subseteq R$ , then either

- $(X \rightarrow Y)$  is trivial, or
- *X* is a superkey of *R*, or
- each attribute of *Y* contained in a candidate key of *R*

A schema  $\{R_1, \dots, R_n\}$  is in 3NF if each relation schema *R<sub>i</sub>* is in 3NF.

- 3NF is looser than BCNF  
⇒ allows more redundancy  
⇒  $R = \{A, B, C\}$  and  $F = \{AB \rightarrow C, C \rightarrow B\}$ .
- lossless-join, dependency-preserving decomposition into 3NF relation schemas always exists.

## Minimal Cover

**Definition:** Two sets of dependencies  $F$  and  $G$  are **equivalent** iff  $F^+ = G^+$ .

There are different sets of functional dependencies that have the same logical implications. Simple sets are desirable.

**Definition:** A set of dependencies  $G$  is **minimal** if

- 1 every right-hand side of a dependency in  $F$  is a single attribute.
- 2 for no  $X \rightarrow A$  is the set  $F - \{X \rightarrow A\}$  equivalent to  $F$ .
- 3 for no  $X \rightarrow A$  and  $Z$  a proper subset of  $X$  is the set  $F - \{X \rightarrow A\} \cup \{Z \rightarrow A\}$  equivalent to  $F$ .

**Theorem:** For every set of dependencies  $F$  there is an equivalent minimal set of dependencies (**minimal cover**).

## Computing a 3NF Decomposition

A lossless-join 3NF decomposition that is dependency preserving can be efficiently computed

```
function Compute3NF( $R, F$ )  
begin  
   $Result := \emptyset$ ;  
   $F' :=$  a minimal cover for  $F$ ;  
  for each  $(X \rightarrow Y) \in F'$  do  
     $Result := Result \cup \{XY\}$ ;  
  if there is no  $R_i \in Result$  such that  
     $R_i$  contains a candidate key for  $R$  then begin  
    compute a candidate key  $K$  for  $R$ ;  
     $Result := Result \cup \{K\}$ ;  
  end;  
  return  $Result$ ;  
end
```

## Finding Minimal Covers

A minimal cover for  $F$  can be computed in four steps. Note that each step must be repeated until it no longer succeeds in updating  $F$ .

### Step 1.

Replace  $X \rightarrow YZ$  with the pair  $X \rightarrow Y$  and  $X \rightarrow Z$ .

### Step 2.

Remove  $X \rightarrow A$  from  $F$  if  $A \in \text{Compute}X^+(X, F - \{X \rightarrow A\})$ .

### Step 3.

Remove  $A$  from the left-hand-side of  $X \rightarrow B$  in  $F$  if  
 $B$  is in  $\text{Compute}X^+(X - \{A\}, F)$ .

### Step 4.

Replace  $X \rightarrow Y$  and  $X \rightarrow Z$  in  $F$  by  $X \rightarrow YZ$ .

## Summary

- functional dependencies provide clues towards elimination of (some) *redundancies* in a relational schema.
- Goals: to decompose relational schemas in such a way that the decomposition is
  - (1) lossless-join
  - (2) dependency preserving
  - (3) BCNF (and if we fail here, at least 3NF)