4.Normalization: Quality of relational designs

4.1 Functional Dependencies

- 4.1.1 Design quality
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- 4.3 Algorithms for finding Normal Forms
- 4.4 Normal Forms: Critical review
- Lit: Kemper/Eickler: chap 6; Garcia-Molina/Ullman/Widom: chap 3.4 ff.; Elmasr/Navathe: chap 14, Kifer et al.: chap. 6

4.1.1 Design quality



What is a "good" conceptual model ?

Usually many alternatives. No clear guidelines, best practice.

Wanted: Formal methods for comparing designs

But...

Use common sense! Simple problems have simple solutions! "Design is an art but a science" Informal guidelines (1)



Avoid redundancies:

Example:			
CREATE TABLE employee (
p#	serial,		
name	VARCHAR(30),		
• • •			
qualification	VARCHAR(20),	typicallymore	
		than one	
PRIMARY KEY (p#, qualification));			
More than one qualification, primitive values			
$\Rightarrow \mathbf{p}$ # is not a key. W	· •		
	<u>p</u> #	name <u>qual</u>	
Dumb design!	22	Meyer programmer	
	27	Müller secretary	
	22	Meyer DB admin	

Informal guidelines (2)



Avoid modeling more than one object from reality in one entity / relation

CREATE	TABLE Experiment (
	id	SERIAL PRIMARY KEY,	
	responsible_Scientist VARCHAR(40),		
	institute	VARCHAR (30),	
	phone	INT,	
	purpose	VARCHAR(100),	
	start	TIMESTAMP,	
	endTime	TIMESTAMP,	
	result	INT)	

What is the problem with this table design?

Informal guidelines (2)



Avoid modeling more than one object from reality in one entity / relation

CREATE	TABLE Experiment (
	id SERIAL PRIMARY KEY,	
	responsible_Scientist VARCHAR(40),	
	institute	VARCHAR (30),
	phone	INT,
	purpose	VARCHAR(100),
	start	TIMESTAMP,
	endTime	TIMESTAMP,
	result	INT)

Redundancy





Redundancies may cause "anomalies"

Deletion of a row may delete all data about a different object

Update of an attribute may cause update on many rows

Insertion may be difficult / impossible, since data are missing

"update anomaly": deletion, update or insertion anomaly

Update anomalies: Examples



CREATE TABLE Experiment (id SERIAL PRIMARY KEY, responsible_Scientist VARCHAR(40), institute VARCHAR (30), phone INT, purpose VARCHAR(100), Table definition using start TIMESTAMP, SQL / DDL endTime TIMESTAMP, (PostgresSQL) result INT)

DELETE FROM Experiment WHERE result < 10

Data on experimenter may be lost!

Update anomaly



Example

update Experiment set phone = 37784
where responsible_Scientist

```
= 'Müller-Lüdenscheid'
```

all those experiment tuples the experimentor of which was 'Müller-Lüdenscheid' have to be changed

⇒ update anomaly

What is an insertion anomaly?

4.1.3 Functions, Functional dependencies rsität

Important formal concept: Functions Functional dependencies (FD) generalize key concept

FDs are used to **formalize integrity constraints** on attributes and relationships Remember **keys** and **1:N relationships**.

If *k* is a key of relation *R(k,a,b,...c)* and *k_val* a key-value, what can be said about the attribute values?

Next to trivial: values are unique for a given key!

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```
Example:
Experiment (id, responsible_Scientist,
institute, phone,...result)
```

"The same responsible_person cannot have different affiliations (institutes)"

{responsible_Scientist} -> {institute}
is a function which expresses the constraint:

Formal notation of RDM



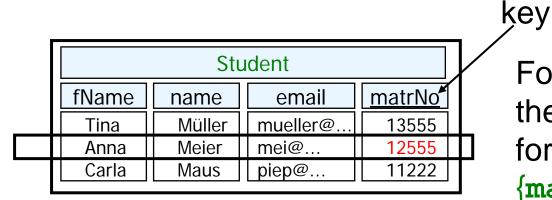
Relation: $R \subseteq dom(a1) \times dom(a2) \times ... \times dom(an)$ **Attribute set**: $\Sigma(R) = \{a1, a2, ..., an\} = R_A$, signature **Tuple**: $r \in R$ **Degree** of R: number of attributes **Relation Schema**: R(a1, a2, ..., an) Database **schema**: set of relation schemas different notations in use

Terminology: Relation = table (file) but relations are sets. tables may have duplicate entries Tuple = row, record Attribute = field (component) Functional Dependencies and keys Freie Universität

Property of a key : at most one row for each value

Given a key, i.e. values for the key attributes, then the value of each $a \in \Sigma(R)$ is unique or all values are undefined.

 \Rightarrow each subset of $\Sigma(R)$ is functional dependent on K



For a given key value, there is a unique value for each attribute e.g. $\{matrNo\} \rightarrow \{fName\}$

 \Rightarrow





Functional and **key dependencies** are **constraints** (invariants) of the application domain

"Functional dependency" constraints *have to be identified during requirements engineering* – like all constraints.

Ultimate goal: DBS monitors compliance with DB state.

Example: Experiment (id, responsible_Scientist, institute, phone,...result)

What has to be done, when a new experiment is inserted?

Motivation for Normal Forms



Suppose we can find a relational schema which **has only keyinduced functional dependencies** (FD) (and "trivial" ones like {a,b} -> {b})

How can we efficiently check the DB state after an update with respect to FD? Do they still hold?

A "good" schema avoids

- Update anomalies
- Costly check of functional dependencies after update

Functional Dependency: Definition Freie Universität

Def.: Functional Dependencies (FDs) Let A = $\Sigma(R)^* = \{a, b, c, ..., a_i, ...\}$ be the attribute set of a relation Rand X, Y ⊆ A, r, r' ∈ R, r ≠ r' **Y is functionally dependent on X** (written: **X** → **Y**) :⇔ (∀ xi ∈ X) r.xi = r'.xi ⇒ (∀ yi ∈ Y) r.yi = r'.yi

- As we know: invariants are independent of the particular database state
- They must hold at all times, i.e. they **restrict the valid states of the database**





Trivial functional dependency

 $X \subseteq Y \Rightarrow Y \to X$

Augmentation

$$\mathsf{Z} \subseteq \mathsf{A} = \Sigma(\mathsf{R}), \quad \mathsf{X} \to \mathsf{Y} \Longrightarrow \mathsf{X} \mathsf{Z} \to \mathsf{Y} \mathsf{Z}$$

Transitivity

$$X,Y,Z \subseteq A=\Sigma(R), X \to Y, Y \to Z \implies X \to Z$$

Proof?

Notation XY -> Z means $X \cup Y$ -> Z





A functional dependency $Y \rightarrow Z$ is called <u>implied</u> by a set F= {F1, ..., Fn} of functional dependencies, if $Y \rightarrow Z$ can be **proven from F.**

A functional dependency $Y \rightarrow Z$ can be <u>inferred</u> (\vdash) by a set of inference rules $R = \{r1, ..., rm\}$ from set $F = \{F1, ..., Fn\}$ of functional dependencies if $Y \rightarrow Z$ can be constructed by a finite number of **syntactic** transformations of F according to rules ri Armstrong inference rules



Given a set of FDs, find all implied FD's

A sound, complete, minimal set (Armstrong axioms):

 $Y \subseteq X \vdash X \rightarrow Y$ (I: inclusion) $\{X \rightarrow Y \ , \ Y \rightarrow Z\} \vdash X \rightarrow Z$ (T: transitivity) $\{X \rightarrow Y\} \vdash XZ \rightarrow YZ$ (A: augmentation)

Sound:

Only implied FDs are constructed by the inference rules

Complete:

Every implied FD will be produced by a finite number of inferences





Given

- schema signature $\Sigma(R) = \{a1, ..., an\}$
- Primary key $P = \{p_1, \dots, p_e\}$
- Set of **candidate keys** $C = \{ \{k_1, ..., k_f\}, ..., \{k'_1, ..., k'_a\} \}$

Def.: $K = P \cup (\cup C)$ are called **prime** (or key) **attributes** i.e. attributes belonging to any candidate key. S(R) \ K : **non-prime** (or non-key) attributes





Types of functional dependencies:

- 1. Key dependencies
- 2. Partial dependencies on one of the candidate keys

```
Expl: {p#} -> {name}
```

// R(<u>p#</u>,name,<u>qualification</u>, ...)

since key is {p#,qualification}

3. Dependencies among non-key attributes expl.: {responsible_Scientist} -> {institute}
4. Dependencies among attributes of different candidate keys



Roadmap

Functional dependencies may cause "update anomalies" ☑ Update anomalies cause troubles

 $\Rightarrow\,$ find relational schema without "anomalies" in case of update \blacksquare

Define "Normal forms" for relations which do not show (all) anomalies

Given a set of functional dependencies, find algorithm which generates a relational schema in some normal form





4.2.1 Definitions

Def.: First normal form A relation is in 1NF :⇔ all attributes are single valued and atomic

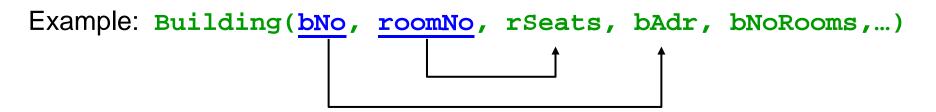
Example: Customer (<u>c_id</u>, name, ..., {phone}, ...) (53,'Miller', ..., {47653, 478992},...)

Equivalent to *Key dependency property*: every attribute is functionally dependent on e very candidate key Second Normal Form (2NF)



Def.: R is in **Second Normal Form (2NF)**, : \Leftrightarrow $\forall X \subseteq \Sigma (R), \forall a \in \Sigma (R)$: $a \notin X \land a$ is not a prime attribute $\land X \rightarrow a$ \Rightarrow X is a key or a superset of a key but not a proper subset of any key of R

This <u>means</u>: **No non-prime** attribute **functionally depends** on only **a part of a key** ("<u>No partial dependency</u>")



Third Normal Form (3NF)



Def.: R is in **third normal form** (3NF) : $\forall X \subseteq \Sigma (R), \forall a \in \Sigma (R) : a \notin X \land X \rightarrow a$ \Rightarrow X contains a key or a is prime

\Leftrightarrow

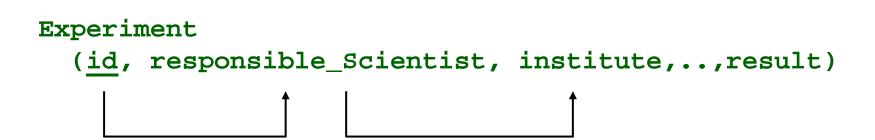
There is **no functional dependency between non-prime attributes** (or attribute sets). Proof?

More on 3NF



Equivalent definition: R is in 3NF :⇔ no non-prime attribute depends transitively on a key.

A non-prime attribute y is **transitively dependent** on a key K, if $K \rightarrow X$ and $X \rightarrow y$ and *not* $X \rightarrow K$ Notation: $K \rightarrow X \rightarrow y$







R is in 3NF \Rightarrow R is in 2NF

Proof:

- Suppose R not in 2NF
 - \Rightarrow exists candidate key K and X \subset K and non-prime attribute
 - a and $X \rightarrow a$ (i.e. exists partial dependency)

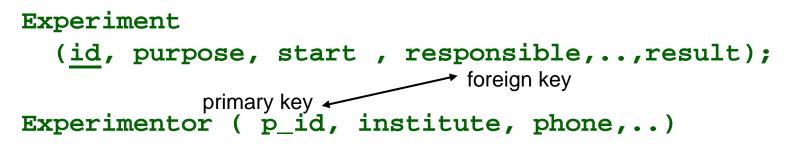
Since K is a key $\Rightarrow K \rightarrow X \rightarrow a$, contradiction to 3NF



(<u>id</u>, responsible_Scientist, institute,..,result)

Indicates the modeling of two different 'real world entities' as one relation

Split into two relations:



DESIGN QUALITY: what do we have?e Universität

- 1. Key dependencies \leftarrow 1NF
- 2. No partial dependencies on candidate keys \leftarrow 2NF
- 4. Dependencies among attributes of different candidate keys ??

⇒ there is at least one more Normal Form which excludes FDs between prime attributes.

Berlin





Dependencies among key attributes

There are relations in 3NF with nontrivial functional dependencies!

Example (*): R(p, o, s, n) with keys {o,s,n} and {p,s,n}, FD p -> o

R in 3NF, but transitive dependency involving key attribute o: {p,s,n} -> p -> o

(*) Interpretation e.g.: PLZ, Ortsteil, Straße, Nummer (in Germany) Annahme: (Ort, Straße, Nr), (PLZ, Straße, Nr) eindeutig

Beyond 3NF



Characterizing essential (= nontrivial and no superkey involved) FDs in 3NF relations

Proposition:

If R is in 3NF and X is a *proper subset* of a candidate key K and $X \rightarrow a$ for some attribute a $\Rightarrow a \notin K \land \exists key K': a \in K'$

Corollary: If $X \rightarrow a$ is an essential FD of a relation in 3NF, and X not a superkey

 \Rightarrow X and a belong to different candidate keys



Def.: A relation R is in **BCNF** : \Leftrightarrow if there is a non-trivial dependencies X \rightarrow a then X is a superkey of R

Equivalent to:

 $X \rightarrow a \Rightarrow$ (i) trivial or (ii) X contains a key of R (iii) There a no essential FDs in R Obvious consequence: **BCNF** \Rightarrow **3NF**





3NF more important in practice than BCNF

Partial dependencies of candidate keys infrequent R relation in 3NF and candidate keys have only one attribute each \Rightarrow R is in BCNF

R in 3 NF and at most one candidate key has more than one attribute \Rightarrow R is in BCNF

Proof?





Last proposition useful in many practical situation: If a relation R has a multi-attribute key and a unique identifier (e.g. a sequence number) then 3NF implies BCNF

e.g. Customer (cID, name, city, street, no, discount, ...)
has keys {cID} and {name, city, street, no}

4.2.2 Lossless property and preserving dependencies



Normalization (by decomposition)

Given relation R having schema $\Sigma(R)$ and FD = {X \rightarrow Y | X,Y $\subseteq \Sigma(R)$ } set of FDs,

Find a set R_1 , ..., R_n of relations in 3NF / BCNF such that:

- $\Sigma(\mathsf{R}) = \bigcup \Sigma(\mathsf{R}_i)$
- For each f = X \rightarrow Y \in FD there exists R_i such that X \cup Y $\subseteq \Sigma(R_i)$

"Dependency preserving"

• R can be reconstructed from R_i, i=1..n



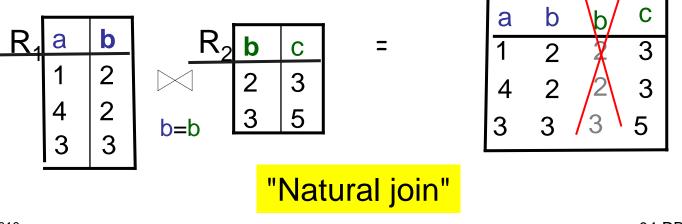
Joining relations



When relation R has been split into relations $R_1, R_2, ..., R_n$, reconstruction of R from $R_1, ..., R_n$ by means of the **join operator**

Join operation (natural join):

concatenate those tuples of R and S which have same name and same value. Eliminate the redundant attribute.

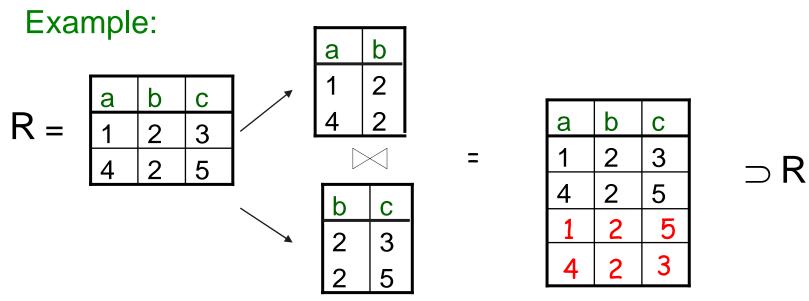






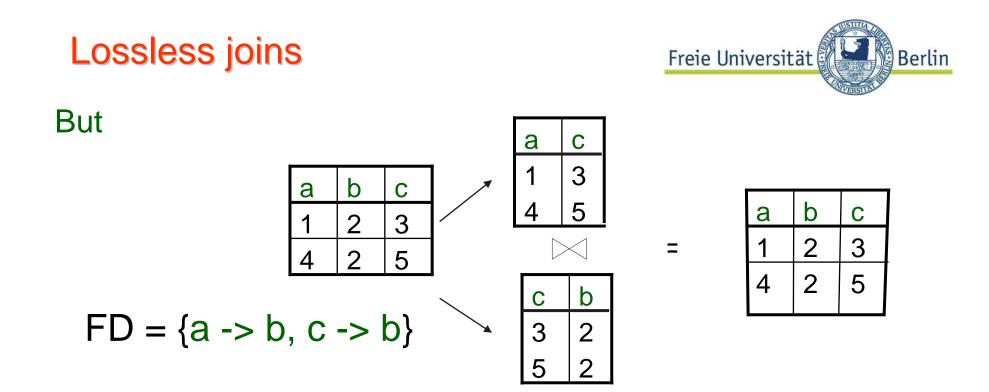
Criterion for "preserved information (losslessness)":

 $R_1 \bowtie R_2 \bowtie \ldots \bowtie R_n = R$



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In general:

Decomposition of R into R1 and R2 is **lossless**, if $\Sigma(R1) \cap \Sigma(R2) \rightarrow \Sigma(R2)$ or $\Sigma(R1) \cap \Sigma(R2) \rightarrow \Sigma(R1)$





Lossless decomposition and keys $\Sigma(R1) \cap \Sigma(R2) \rightarrow \Sigma(R2)$ or $\Sigma(R1) \cap \Sigma(R2) \rightarrow \Sigma(R1)$

The common attribute(s) of R1 and R2 are a key (or a superset of a key) of R1 or R2

Functional dependencies are transformed into key dependencies

 Invariance property expressed by FDs may now be checked by checking the primary key property - efficiently done by every DBS

BCNF and 3NF



BCNF does not always guarantees both the lossless property and dependency preservation

Example: R(p, o, s, n) with keys {o,s,n} and {p,s,n}, FD p -> o Normalisation to BCNF: R1 (p,s,n) and R2(p,o) ⇒ Dependency (o,s,n) -> p is lost

Consequence: Normalization to 3NF is the best to achieve in general

4.2.5 Multivalued dependencies and ANFiversität

Example

Hobbies(<u>name,affiliation,hobby</u>)

Meier Müller Meier	FUB TUB FUB	skiing trekking trekking	can have one or more affiliations, and one or more hobbies, e.g
Schulze	HU	skating	Schulze {HU.FU} {tennis,skating}
Schulze	FU	tennis	
Schulze	FU	skating	These rows must
Schulze	HU	tennis	exist (MVD restriction)

- Two multivalued attributes: affiliation, hobbies, both dependent on name.
- Introduce redundancy
- MVD defines which tuples must exist.

Assumption: a person





Example with MV attributes:



MVD are invariants on the existence of tuples if multivalued attributes are represented in 1NF

```
Def.: MVD (multivalued dependency)
Let R = (a, y, b),
b is multivalued dependent on a (a ->>b)
if for each value v of a
\{v\} \times (\pi_v(\sigma_{a=v} R)) \times (\pi_b(\sigma_{a=v} R)) \subseteq R
```

MVD: example



```
Example:

{'Meier'} \times {'FU'} \times {'skiing', 'trekking'} \subseteq Person

{'Müller'} \times {'TU'} \times {'trekking'} \subseteq Person

{'Schulze'} \times {'HU',FU} \times {'skating', 'tennis'} \subseteq Person
```

Meier	FUB	skiing
Müller	TUB	trekking
Meier	FUB	trekking
Schulze	HU	skating
Schulze	FU	tennis
Schulze	FU	skating
Schulze	HU	tennis

'hobby' is mv-dependent on 'name': **name** ->> **hobby**

Fourth Normal Form



Def.: Let A, $B \subseteq \Sigma(R)$; R is in **Fourth Normal Form** if for every MVD A ->> B (i) $B \subseteq A$ or (ii) $B = \Sigma(R) \setminus A$ or (iii) A contains a key

Example not in 4NF, check

Normalized representation:

Müller	TUB
Meier	FUB
Schulze	HU
Schulze	FU

Müller	trekking	
Meier	trekking	
Meier	skiing	
Schulze	skating	
Schulze	tennis	

Normal forms: summary



Normal forms are quality criteria for database design. <u>Important:</u> 1NF – 3NF Exotics: BCNF, 4NF (and higher!)

2NF / 3NF formalize the basic design principle: **"Never mix up different real world entities into a single design object (e.g. entity)**"

2NF / 3NF already defined for ERM, since FDs are given (result of requirement analysis, just like key dependencies). 4.3 Finding Normal Forms



Invariants of application domain have to be made explicit during requirements analysis

e.g. "A scientist has at most one affiliation – her institute"

"A region-id is unique within a country"

"A person has exactly one date of birth"

Formalization ⇒ Functional Dependencies Wanted: algorithm producing "good" relational schemas from the set DEP of all FDs

FDs and Normal Forms



Given a set of dependencies DEP there are two approaches:

• Synthesis

Set up relations in such a way, that

- All attributes are consumed
- The relations are in normal form

• Decomposition

For a given set of relations find those which are not normalized with respect to DEP and decompose them into normalized relations **Decomposition: eliminate FDs**



Given $\Sigma(R) = U$ and DEP the set of FDs **Algorithm DECOMP(R)**:

(i) Find the set of keys K: $K \rightarrow U \in DEP \text{ or } K \rightarrow U \in DEP^+$ (DEP+ set of all implied dependencies)

(ii) Eliminate all transitive dependencies by splitting recursively: {if K → Y -> a is a transitive FD in R_k, split R_k into R_i, R_j Σ(R_i) = Σ(R_k) \ {a}, Σ(R_j) = Y ∪ {a} }
(iii) If no more relations R_k with transitive dependency exit else for all R_k DECOMP(R_k)

Synthesis



Disadvantage of decomposition: inefficient (e.g. determination of keys) produces more relations than necessary

Synthesis

Given relation R and set of FDs DEP

Find a canonical set MIN of FDs which "covers" DEP and is minimal.

Construct normalized Relations R_k from MIN with $\bigcup \Sigma(R_k) = \Sigma(R)$

Finding a canonical set of FDs



Given a set of FDs DEP and a relational schema R

- Find a minimal set **MIN** such that $DEP \subseteq MIN^+$
- Find a relational schema in 3NF, from which R can be losslessly reconstructed

MIN is called minimal cover of DEP

Definitions

 $\begin{array}{l} X \rightarrow Y \in \textbf{MIN}^{+} :\Leftrightarrow X \rightarrow Y \ \text{ can be proven from} \\ \text{ the FDs } \in \text{MIN} \\ \text{ e.g. } \{a \rightarrow b, \, b \rightarrow c\}^{+} = \{a \rightarrow b, \, b \rightarrow c, \, a \rightarrow c \,\} \\ \text{MIN is minimal} :\Leftrightarrow \text{ for every FD f} \ (\text{MIN} \setminus f)^{+} \neq \text{MIN}^{+} \end{array}$

Finding a minimal cover (1)



Example:

 $ab \rightarrow d, b \rightarrow c, dc \rightarrow e \}+ = \{ab \rightarrow d, b \rightarrow c, a \rightarrow dc, dc \rightarrow e, ab \rightarrow e\} \}$ Important first step:

Given a set of attributes X, determine all attributes (closure of X) which can be functionally determined by X?

Def.: **Closure of attribute set X** with respect to the set DEP of FDs is the largest set Y of attributes such that $X \rightarrow Y \in DEP^+$

Closure of attribute set X



```
I = 0; X[0] = X; /* integer I, attr. set X[0] */

REPEAT /* loop to find larger X[I] */

I = I + 1; /* new I */

X[I] = X[I-1]; /* initialize new X[I] */

FOR ALL Z->W in DEP /* loop on all FDs Z ->W in DEP*/

IF Z \subseteqX[I] /* if Z contained in X[I] */

THEN X[I] = X[I]\cup W; /* add attributes in W to X[I]*/

END FOR /* end loop on FDs */

UNTIL X[I] = X[I-1]; /* loop till no new attributes*/

RETURN X = X[I]; /* return closure of X */
```

Rule used: X -> YZ and Z -> W then X -> YZW Proof?

Finding a canonical set



Algorithm for determining a <u>minimal cover</u> in polynomial time

Step 1: Normalization

Replace each FD X \rightarrow Y of DEP in which Y contains more than one attribute, by FDs with one attribute on the right hand side

Example:

 $\mathsf{DEP} = \{\mathsf{ab} \rightarrow \mathsf{cd}, \mathsf{a} \rightarrow \mathsf{e}\} \rightarrow \{\mathsf{ab} \rightarrow \mathsf{c}, \mathsf{ab} \rightarrow \mathsf{d}, \mathsf{a} \rightarrow \mathsf{e}\}$

Finding a canonical set (2)



Step 2: Remove redundant FDs

f is redundant, if (DEP \ {f})⁺ = DEP⁺ For each f = X \rightarrow Y \in DEP : if Y \subseteq (X+) using only FDs \in DEP \{f} then f is redundant else not redundant Example: {b \rightarrow d, d \rightarrow e, ef \rightarrow a, c \rightarrow f, bc \rightarrow a} bc+ = {b,d,e,f,a} \supseteq {a} \Rightarrow FD f= bc -> a is redundant Explicit derivation not using f: b -> d, d -> e \Rightarrow b -> e, c -> f \Rightarrow bc -> ef , ef -> a \Rightarrow bc -> a Finding a canonical set (3)



Step 3: Minimize left hand side

For of each FD f = X
$$\rightarrow$$
 Y \in DEP, a \in X
DEP'= {DEP \ f } \cup { X \ {a} \rightarrow Y}
if {DEP}+ = (DEP')+
then replace DEP by DEP'

end_for

If a FD f has been minimized repeat step 2

Example:

{bcd -> a, c -> e, e -> b} + = {cd -> a, c -> e, e -> b} +

Finding a canonical set (4)



Step 4: Unify left hand side

Applying the union rule " X \rightarrow Y \land X \rightarrow Z \Rightarrow X \rightarrow YZ "

Example: {cd -> a, cd -> e, d -> f} unified: {cd -> ae, d -> f}

The remaining set of Functional Dependencies is the minimal cover of the original set DEP of FDs

Synthesis algorithm

Normalization problem:



Given a relation R in 1NF and a set DEP of FD Find a lossless, dependency preserving decomposition R_1, \ldots, R_k , all of them in 3NF

Synthesis Algorithm

Find minimal cover MIN of DEP; For all X -> Y in MIN define a relation RX with schema $\Sigma(RX) = X \cup Y$ Assign all FDs X' -> Y' with X' \cup Y' $\subseteq \Sigma(RX)$ to RX If none of the synthesized relations RX contains a candidate key of R then introduce a relation Rkey which contains a candidate key of R Remove relations RY where: $\Sigma(RY) \subseteq \Sigma(RX)$ **Normal Forms: Critical review**



Should relations be always normalized ?

- Yes : makes invariant checking easy, and no "update anomalies"
- No: Why should we normalize if there are no updates ?
- Example: Customer(cu_ld, name, fname, zipCode, city, street, no) No reason to normalize if only one address per customer and updates are infrequent
- Consider cost of joins / updates
 - How expensive are selects which need joins because of normalization?
 - Updates which cause anomalies?

ER modeling and Normal Forms



ER and Normal Forms:

Two different mechanisms to design a database scheme

ER more intuitive,

NF uses algorithms

ER-models often already in 3NF

Use normalization as a complementary design method

- Set up ER model
- Transform to relations
- Normalize each non normalized relation if the tradeoff of join processing and updating redundant data suggests to do so