## FUNCTIONAL DEPENDENCY THEORY II

CS121: Relational Databases
Fall 2018 - Lecture 20

## Canonical Cover

$\square$ A canonical cover $F_{c}$ for $F$ is a set of functional dependencies such that:
$\square F$ logically implies all dependencies in $F_{c}$
$\square F_{\mathrm{c}}$ logically implies all dependencies in $F$
$\square$ Can't infer any functional dependency in $F_{c}$ from other dependencies in $F_{c}$
$\square$ No functional dependency in $F_{c}$ contains an extraneous attribute
$\square$ Left side of all functional dependencies in $F_{c}$ are unique

- There are no two dependencies $\alpha_{1} \rightarrow \beta_{1}$ and $\alpha_{2} \rightarrow \beta_{2}$ in $F_{c}$ such that $\alpha_{1}=\alpha_{2}$


## Extraneous Attributes

$\square$ Given a set $F$ of functional dependencies
$\square$ An attribute in a functional dependency is extraneous if it can be removed from $F$ without changing $F^{+}$
$\square$ Formally: given $F$, and $\alpha \rightarrow \beta$

- If $A \in \alpha$, and $F$ logically implies $(F-\{\alpha \rightarrow \beta\}) \cup\{(\alpha-A) \rightarrow \beta\}$, then $A$ is extraneous
$\square$ If $A \in \beta$, and $(F-\{\alpha \rightarrow \beta\}) \cup\{\alpha \rightarrow(\beta-A)\}$ logically implies $F$, then $A$ is extraneous
- i.e. generate a new set of functional dependencies $F^{\prime}$ by replacing $\alpha \rightarrow \beta$ with $\alpha \rightarrow(\beta-A)$
- See if $F^{\prime}$ logically implies $F$


## Testing Extraneous Attributes

$\square$ Given relation schema $R$, and a set $F$ of functional dependencies that hold on $R$
$\square$ Attribute $A$ in $\alpha \rightarrow \beta$
$\square$ If $A \in \alpha$ (i.e. $A$ is on left side of the dependency), then let $\gamma=\alpha-\{A\}$
$\square$ See if $\gamma \rightarrow \beta$ can be inferred from $F$
$\square$ Compute $\gamma^{+}$under $F$

- If $\beta \subseteq \gamma^{+}$then $A$ is extraneous in $\alpha$
$\square$ e.g. if $A B \rightarrow C$ and you want to see if $B$ is extraneous, can see if you can infer $A \rightarrow C$ from $F$


## Testing Extraneous Attributes (2)

$\square$ Given relation schema $R$, and a set $F$ of functional dependencies that hold on $R$
$\square$ Attribute $A$ in $\alpha \rightarrow \beta$
$\square$ If $A \in \beta$ (on right side of the dependency), then try the altered set $F^{\prime}$

- $F^{\prime}=(F-\{\alpha \rightarrow \beta\}) \cup\{\alpha \rightarrow(\beta-A)\}$
$\square$ See if $\alpha \rightarrow A$ can be inferred from $F^{\prime}$
$\square$ Compute $\alpha^{+}$under $F^{\prime}$
- If $\alpha^{+}$includes $A$ then $A$ is extraneous in $\beta$
$\square$ e.g. if $A \rightarrow B C$ and you want to see if $B$ is extraneous, you can already infer $A \rightarrow B$ from this dependency
$\square$ Must generate $F^{\prime}$ with only $A \rightarrow C$, and if you can infer $A \rightarrow B$ from $F^{\prime}$, then $B$ was indeed extraneous


## Computing Canonical Cover

$\square$ A simple way to compute the canonical cover of $F$
$F_{\mathrm{c}}=F$
repeat
apply union rule to replace dependencies in $F_{c}$ of form
$\alpha_{1} \rightarrow \beta_{1}$ and $\alpha_{1} \rightarrow \beta_{2}$ with $\alpha_{1} \rightarrow \beta_{1} \beta_{2}$
find a functional dependency $\alpha \rightarrow \beta$ in $F_{c}$ with an extraneous attribute
/* Use $F_{c}$ for the extraneous attribute test, not $F!!!$ */
if an extraneous attribute is found, delete it from $\alpha \rightarrow \beta$
until $F_{c}$ stops changing

## Canonical Cover Example

$\square$ Functional dependencies $F$ on schema ( $A, B, C$ )
$\square F=\{A \rightarrow B C, B \rightarrow C, A \rightarrow B, A B \rightarrow C\}$
$\square$ Find $F_{c}$
$\square$ Apply union rule to $A \rightarrow B C$ and $A \rightarrow B$
$\square$ Left with: $\{A \rightarrow B C, B \rightarrow C, A B \rightarrow C$ \}
$\square A$ is extraneous in $A B \rightarrow C$
$\square B \rightarrow C$ is logically implied by $F$ (obvious)
$\square$ Left with: $\{A \rightarrow B C, B \rightarrow C$ \}
$\square C$ is extraneous in $A \rightarrow B C$
Logically implied by $A \rightarrow B, B \rightarrow C$
$\square F_{c}=\{A \rightarrow B, B \rightarrow C\}$

## Canonical Covers

$\square$ A set of functional dependencies can have multiple canonical covers
$\square$ Example:
$\square F=\{A \rightarrow B C, B \rightarrow A C, C \rightarrow A B\}$
$\square$ Has several canonical covers:
$\square F_{c}=\{A \rightarrow B, B \rightarrow C, C \rightarrow A\}$
$\square F_{c}=\{A \rightarrow B, B \rightarrow A C, C \rightarrow B\}$
$\square F_{c}=\{A \rightarrow C, C \rightarrow B, B \rightarrow A\}$
$\square F_{\mathrm{c}}=\{A \rightarrow C, B \rightarrow C, C \rightarrow A B\}$
$\square F_{c}=\{A \rightarrow B C, B \rightarrow A, C \rightarrow A\}$

## Another Example

$\square$ Functional dependencies $F$ on schema ( $A, B, C, D$ )
$\square F=\{A \rightarrow B, B C \rightarrow D, A C \rightarrow D\}$
$\square$ Find $F_{c}$
$\square$ In this case, it may look like $F_{c}=F \ldots$
$\square$ However, can infer $A C \rightarrow D$ from $A \rightarrow B, B C \rightarrow D$ (pseudotransitivity), so $A C \rightarrow D$ is extraneous in $F$
$\square$ Therefore, $F_{c}=\{A \rightarrow B, B C \rightarrow D\}$
$\square$ Alternately, can argue that $D$ is extraneous in $A C \rightarrow D$
$\square$ With $F^{\prime}=\{A \rightarrow B, B C \rightarrow D\}$, we see that $\{A C\}^{+}=A B C D$, so $D$ is extraneous in $A C \rightarrow D$
$\square$ (If you eliminate the entire RHS of a functional dependency, it goes away)

## Lossy Decompositions

$\square$ Some schema decompositions lose information
$\square$ Example:
employee(emp id, emp_name, phone, title, salary, start_date)
$\square$ Decomposed into:
emp_ids(emp id, emp_name)
emp_details(emp_name, phone, title, salary, start_date)
$\square$ Problem:
■emp_name doesn't uniquely identify employees
$\square$ This is a lossy decomposition

## Lossless Decompositions

$\square$ Given:
$\square$ Relation schema $R$, relation $r(R)$
$\square$ Set of functional dependencies $F$
$\square$ Let $R_{1}$ and $R_{2}$ be a decomposition of $R$
$\square R_{1} \cup R_{2}=R$
$\square$ The decomposition is lossless if, for all legal instances of $r$ :
$\Pi_{R_{1}}(r) \bowtie \Pi_{R_{2}}(r)=r$
$\square$ A simple definition...

## Lossless Decompositions (2)

$\square$ Can define with functional dependencies:
$\square R_{1}$ and $R_{2}$ form a lossless decomposition of $R$ if at least one of these dependencies is in $F^{+}$:

$$
\begin{aligned}
& R_{1} \cap R_{2} \rightarrow R_{1} \\
& R_{1} \cap R_{2} \rightarrow R_{2}
\end{aligned}
$$

$\square R_{1} \cap R_{2}$ forms a superkey of $R_{1}$ and/or $R_{2}$
$\square$ Test for superkeys using attribute-set closure

## Decomposition Examples (1)

$\square$ The employee example:
employee(emp id, emp_name, phone, title, salary, start_date)
$\square$ Decomposed into:
emp_ids(emp id, emp_name)
emp_details(emp_name, phone, title, salary, start_date)
$\square$ emp_name is not a superkey of emp_ids or emp_details, so the decomposition is lossy

## Decomposition Examples (2)

$\square$ The bor_loan example:
bor_loan(cust id, loan id, amount)
$\square$ Decomposed into:
borrower(cust_id, loan_id)
loan(loan id, amount) (loan_id $\rightarrow$ loan_id, amount )
$\square$ loan_id is a superkey of loan, so the decomposition is lossless

## BCNF Decompositions

$\square$ If $R$ is a schema not in BCNF:
$\square$ There is at least one nontrivial functional dependency $\alpha \rightarrow \beta$ such that $\alpha$ is not a superkey for $R$

- For simplicity, also require that $\alpha \cap \beta=\varnothing$
- (if $\alpha \cap \beta \neq \emptyset$ then ( $\alpha \cap \beta$ ) is extraneous in $\beta$ )
$\square$ Replace $R$ with two schemas:

$$
\begin{aligned}
R_{1} & =(\alpha \cup \beta) \\
R_{2} & =(R-\beta) \\
& =(\text { was } R-(\beta-\alpha), \text { but } \beta-\alpha=\beta, \text { since } \alpha \cap \beta=\emptyset)
\end{aligned}
$$

$\square$ BCNF decomposition is lossless
$\square R_{1} \cap R_{2}=\alpha$
$\square \alpha$ is a superkey of $R_{1}$
$\square \alpha$ also appears in $R_{2}$

## Dependency Preservation

$\square$ Some schema decompositions are not dependencypreserving
$\square$ Functional dependencies that span multiple relation schemas are hard to enforce
$\square$ e.g. BCNF may require decomposition of a schema for one dependency, and make it hard to enforce another dependency
$\square$ Can test for dependency preservation using functional dependency theory

## Dependency Preservation (2)

$\square$ Given:
$\square$ A set $F$ of functional dependencies on a schema $R$
$\square R_{1}, R_{2}, \ldots, R_{n}$ are a decomposition of $R$
$\square$ The restriction of $F$ to $R_{i}$ is the set $F_{i}$ of functional dependencies in $F^{+}$that only has attributes in $R_{i}$
$\square$ Each $F_{i}$ contains functional dependencies that can be checked efficiently, using only $R_{i}$
$\square$ Find all functional dependencies that can be checked efficiently
$\square F^{\prime}=F_{1} \cup F_{2} \cup \ldots \cup F_{n}$
If $F^{\prime+}=F^{+}$then the decomposition is dependencypreserving

## Third Normal Form Schemas

$\square$ Can generate a 3NF schema from a set of functional dependencies $F$
$\square$ Called the 3NF synthesis algorithm
$\square$ Instead of decomposing an initial schema, generates schemas from a set of dependencies
$\square$ Given a set $F$ of functional dependencies
$\square$ Uses the canonical cover $F_{c}$
$\square$ Ensures that resulting schemas are dependency-preserving

## 3NF Synthesis Algorithm

$\square$ Inputs: set of functional dependences $F$, on a schema $R$
let $F_{\mathrm{c}}$ be a canonical cover for $F$;
$i:=0$;
for each functional dependency $\alpha \rightarrow \beta$ in $F_{c}$ do
if none of the schemas $R_{i j} i=1,2, \ldots, i$ contains $(\alpha \cup \beta)$ then $i:=i+1$; $R_{i}:=(\alpha \cup \beta)$
end if
done
if no schema $R_{i} i=1,2, \ldots, i$ contains a candidate key for $R$ then
$i:=i+1$;
$R_{i}:=$ any candidate key for $R$
end if
return $\left(R_{1}, R_{2}, \ldots, R_{i}\right)$

## BCNF vs. 3NF

$\square$ Boyce-Codd Normal Form:
$\square$ Eliminates more redundant information than 3NF
$\square$ Some functional dependencies become expensive to enforce

- The conditions to enforce involve multiple relations
$\square$ Overall, a very desirable normal form!
$\square$ Third Normal Form:
$\square$ All [more] dependencies are [probably] easy to enforce...
$\square$ Allows more redundant information, which must be kept synchronized by the database application!
$\square$ Personal banker example:
works_in(emp id, branch_name) cust_banker_branch(cust id, branch name, emp_id, type)
- Branch names must be kept synchronized between these relations!


## BCNF and 3NF vs. SQL

$\square$ SQL constraints:
$\square$ Only key constraints are fast and easy to enforce!
$\square$ Only easy to enforce functional dependencies $\alpha \rightarrow \beta$ if $\alpha$ is a key on some table!
$\square$ Other functional dependencies (even "easy" ones in 3NF) may require more expensive constraints, e.g. CHECK
$\square$ For SQL databases with materialized views:
$\square$ Can decompose a schema into BCNF
$\square$ For dependencies $\alpha \rightarrow \beta$ not preserved in decomposition, create materialized view joining all relations in dependency
$\square$ Enforce unique( $\alpha$ ) constraint on materialized view
$\square$ Impacts both space and performance, but it works...

## Multivalued Attributes

$\square \mathrm{E}-\mathrm{R}$ schemas can have multivalued attributes
$\square$ 1NF requires only atomic attributes
$\square$ Not a problem; translating to relational model leaves everything atomic
$\square$ Employee example: employee(emp id, emp_name) emp_deps(emp_id, dependent) emp_nums(emp_id, phone_num)

| employee |
| :--- |
| emp id |
| emp_name |
| $\{$ phone_num $\}$ |
| $\{$ dependent $\}$ |

$\square$ What are the requirements on these schemas for what tuples must appear?

## Multivalued Attributes (2)

$\square$ Example data:

| emp_id | emp_name |
| :---: | :---: |
| 125623 | Rick |
| employee |  |


| emp_id | dependent |
| :---: | :---: |
| 125623 | Jeff |
| 125623 | Alice |
| emp_deps |  |


| emp_id | phone_num |
| :---: | :---: |
| 125623 | $555-8888$ |
| 125623 | $555-2222$ |
| emp_nums |  |

- Every distinct value of multivalued attribute requires a separate tuple, including associated value of emp_id
$\square$ A consequence of 1 NF, in fact!
- If attributes could be nonatomic, could just store list of values in the appropriate column!
$\square 1$ NF requires extra tuples to represent multivalues


## Independent Multivalued Attributes

$\square$ Question is trickier when a schema stores several independent multivalued attributes
$\square$ Proposed combined schema:
employee(emp id, emp_name)
emp_info(emp_id, dependent, phone_num)
$\square$ What tuples must appear in emp_info?

- emp_info is a relation
$\square$ If an employee has $M$ dependents and $N$ phone numbers, emp_info must contain $M \times N$ tuples
- Exactly what we get if we natural-join emp_deps and emp_nums
$\square$ Every combination of the employee's dependents and their phone numbers


## Independent Multivalued Attributes

$\square$ Example data:

| emp_id | emp_name |
| :---: | :---: |
| 125623 | Rick |

employee

| emp_id | dependent | phone_num |
| :---: | :---: | :---: |
| 125623 | Jeff | $555-8888$ |
| 125623 | Jeff | $555-2222$ |
| 125623 | Alice | $555-8888$ |
| 125623 | Alice | $555-2222$ |
| emp_info |  |  |

$\square$ Clearly has unnecessary redundancy
$\square$ Can't formulate functional dependencies to represent multivalued attributes
$\square$ Can't use BCNF or 3NF decompositions to eliminate redundancy in these cases

## Multivalued Attributes Example

$\square$ Two employees: Rick and Bob
$\square$ Both share a phone number at work

- Both have two kids
- Both have a kid named Alice
$\square$ Can't use functional dependencies to reason about this situation!
$\square$ emp_id $\rightarrow$ phone_num doesn't hold since an employee can have several phone numbers
$\square$ phone_num $\rightarrow$ emp_id doesn't hold either, since several employees can have the same phone number
$\square$ Same with emp_id and dependent...

| emp_id | emp_name |
| :---: | :---: |
| 125623 | Rick |
| 127341 | Bob |
| employee |  |


| emp_id | phone_num |
| :---: | :---: |
| 125623 | $555-8888$ |
| 125623 | $555-2222$ |
| 127341 | $555-2222$ |
| emp_nums |  |


| emp_id | dependent |
| :---: | :---: |
| 125623 | Jeff |
| 125623 | Alice |
| 127341 | Alice |
| 127341 | Clara |
| emp_deps |  |

## Dependencies

$\square$ Functional dependencies rule out what tuples can appear in a relation
$\square$ If $A \rightarrow B$ holds, then tuples cannot have same value for $A$ but different values for $B$
$\square$ Also called equality-generating dependencies
$\square$ Multivalued dependencies specify what tuples must be present
$\square$ To represent a multivalued attribute's values properly, a certain set of tuples must be present
$\square$ Also called tuple-generating dependencies

## Multivalued Dependencies

$\square$ Given a relation schema $R$
$\square$ Attribute-sets $\alpha \in R, \beta \in R$
$\square \alpha \rightarrow \beta$ is a multivalued dependency
$\square$ " $\alpha$ multidetermines $\beta$ "
$\square$ A multivalued dependency $\alpha \rightarrow \beta$ holds on $R$ if, in any legal relation $r(R)$ :
For all pairs of tuples $t_{1}$ and $t_{2}$ in $r$ such that $t_{1}[\alpha]=t_{2}[\alpha]$, There also exists tuples $t_{3}$ and $t_{4}$ in $r$ such that:
$\square t_{1}[\alpha]=t_{2}[\alpha]=t_{3}[\alpha]=t_{4}[\alpha]$
$\square t_{1}[\beta]=t_{3}[\beta]$ and $t_{2}[\beta]=t_{4}[\beta]$
$\square t_{1}[R-\beta]=t_{4}[R-\beta]$ and $t_{2}[R-\beta]=t_{3}[R-\beta]$

## Multivalued Dependencies (2)

$\square$ Multivalued dependency $\alpha \rightarrow \beta$ holds on $R$ if, in any legal relation $r(R)$ :
For all pairs of tuples $t_{1}$ and $t_{2}$ in $r$ such that $t_{1}[\alpha]=t_{2}[\alpha]$,
There also exists tuples $t_{3}$ and $t_{4}$ in $r$ such that:
$\square t_{1}[\alpha]=t_{2}[\alpha]=t_{3}[\alpha]=t_{4}[\alpha]$
$\square t_{1}[\beta]=t_{3}[\beta]$ and $t_{2}[\beta]=t_{4}[\beta]$
$\square t_{1}[R-\beta]=t_{4}[R-\beta]$ and $t_{2}[R-\beta]=t_{3}[R-\beta]$
$\square$ Pictorially:

|  | $\alpha$ | $\beta$ | $R-(\alpha \cup \beta)$ |
| :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1} \ldots a_{i}$ | $a_{i+1} \ldots a_{j}$ | $a_{j+1} \ldots a_{n}$ |
| $t_{2}$ | $a_{1} \ldots a_{i}$ | $b_{i+1} \ldots b_{j}$ | $b_{j+1} \ldots b_{n}$ |
| $t_{3}$ | $a_{1} \ldots a_{i}$ | $a_{i+1} \ldots a_{j}$ | $b_{j+1} \ldots b_{n}$ |
| $t_{4}$ | $a_{1} \ldots a_{i}$ | $b_{i+1} \ldots b_{j}$ | $a_{j+1} \ldots a_{n}$ |

## Multivalued Dependencies (3)

$\square$ Multivalued dependency:

|  | $\alpha$ | $\beta$ | $R-(\alpha \cup \beta)$ |
| :---: | :---: | :---: | :---: |
| $t_{1}$ | $a_{1} \ldots a_{i}$ | $a_{i+1} \ldots a_{j}$ | $a_{j+1} \ldots a_{n}$ |
| $t_{2}$ | $a_{1} \ldots a_{i}$ | $b_{i+1} \ldots b_{j}$ | $b_{j+1} \ldots b_{n}$ |
| $t_{3}$ | $a_{1} \ldots a_{i}$ | $a_{i+1} \ldots a_{j}$ | $b_{j+1} \ldots b_{n}$ |
| $t_{4}$ | $a_{1} \ldots a_{i}$ | $b_{i+1} \ldots b_{j}$ | $a_{j+1} \ldots a_{n}$ |

$\square$ If $\alpha \rightarrow \beta$ then $R-(\alpha \cup \beta)$ is independent of this fact

- Every distinct value of $\beta$ must be associated once with every distinct value of $R-(\alpha \cup \beta)$
$\square$ Let $\gamma=R-(\alpha \cup \beta)$
- If $\alpha \rightarrow \beta$ then also $\alpha \rightarrow \gamma$
$\square \alpha \rightarrow \beta$ implies $\alpha \rightarrow \gamma$
$\square$ Sometimes written $\alpha \rightarrow \beta \mid \gamma$


## Trivial Multivalued Dependencies

$\square \alpha \rightarrow \beta$ is a trivial multivalued dependency on $R$ if all relations $r(R)$ satisfy the dependency
$\square$ Specifically, $\alpha \rightarrow \beta$ is trivial if $\beta \subseteq \alpha$, or if
$\alpha \cup \beta=R$
$\square$ Employee examples:
$\square$ For schema emp_deps(emp_id, dependent), emp_id $\rightarrow$ dependent is trivial
$\square$ For emp_info(emp_id, dependent, phone_num), emp_id $\rightarrow$ dependent is not trivial

## Inference Rules

$\square$ Can reason about multivalued dependencies, just like functional dependencies
$\square$ There is a set of complete, sound inference rules for MVDs
$\square$ Example inference rules:
$\square$ Complementation rule:

- If $\alpha \rightarrow \beta$ holds on $R$, then $\alpha \rightarrow R-(\alpha \cup \beta)$ holds
$\square$ Multivalued augmentation rule:
■ If $\alpha \rightarrow \beta$ holds, and $\gamma \subseteq R$, and $\delta \subseteq \gamma$, then $\gamma \alpha \rightarrow \delta \beta$ holds
$\square$ Multivalued transitivity rule:
- If $\alpha \rightarrow \beta$ and $\beta \rightarrow \gamma$ holds, then $\alpha \rightarrow \gamma-\beta$ holds
$\square$ Coalescence rule:
- If $\alpha \rightarrow \beta$ holds, and $\gamma \subseteq \beta$, and there is a $\delta$ such that $\delta \subseteq R$, and $\delta \cap \beta=\emptyset$, and $\delta \rightarrow \gamma$, then $\alpha \rightarrow \gamma$ holds


## Functional Dependencies

$\square$ Functional dependencies are also multivalued dependencies
$\square$ Replication rule:
$\square$ If $\alpha \rightarrow \beta$, then $\alpha \rightarrow \beta$ too
$\square$ Note there is an additional constraint from $\alpha \rightarrow \beta$ : each value of $\alpha$ has at most one associated value for $\beta$
$\square$ Usually, functional dependencies are not stated as multivalued dependencies

- The extra caveat is important, but not obvious in notation
$\square$ Also, functional dependencies are easier to reason about!


## Closures and Restrictions

$\square$ For a set $D$ of functional and multivalued dependencies, can compute closure $D^{+}$
$\square$ Use inference rules for both functional and multivalued dependencies to compute closure
$\square$ Sometimes need the restriction of $D^{+}$to a relation schema $R$, too
$\square$ The restriction of $D$ to a schema $R_{i}$ includes:
$\square$ All functional dependencies in $D^{+}$that include only attributes in $R_{i}$
$\square$ All multivalued dependencies of the form $\alpha \rightarrow \beta \cap R_{i}$, where $\alpha \subseteq R_{i}$, and $\alpha \rightarrow \beta$ is in $D^{+}$

## Fourth Normal Form

$\square$ Given:

- Relation schema $R$
$\square$ Set of functional and multivalued dependencies $D$
$\square R$ is in 4NF with respect to $D$ if:
$\square$ For all multivalued dependencies $\alpha \rightarrow \beta$ in $D^{+}$, where $\alpha \subseteq R$ and $\beta \subseteq R$, at least one of the following holds:
$-\alpha \rightarrow \beta$ is a trivial multivalued dependency
- $\alpha$ is a superkey for $R$
$\square$ Note: If $\alpha \rightarrow \beta$ then $\alpha \rightarrow \beta$
$\square$ A database design is in 4 NF if all schemas in the design are in 4NF


## 4NF and BCNF

$\square$ Main difference between $4 N F$ and BCNF is use of multivalued dependencies instead of functional dependencies
$\square$ Every schema in 4NF is also in BCNF

- If a schema is not in BCNF then there is a nontrivial functional dependency $\alpha \rightarrow \beta$ such that $\alpha$ is not a superkey for $R$
$\square$ If $\alpha \rightarrow \beta$ then $\alpha \rightarrow \beta$


## 4NF Decompositions

$\square$ Decomposition rule is very similar to BCNF
$\square$ If schema $R$ is not in $4 N F$ with respect to a set of multivalued dependencies $D$ :
$\square$ There is some nontrivial dependency $\alpha \rightarrow \beta$ in $D^{+}$ where $\alpha \subseteq R$ and $\beta \subseteq R$, and $\alpha$ is not a superkey of $R$

- Also constrain that $\alpha \cap \beta=\varnothing$
$\square$ Replace $R$ with two new schemas:
$\square R_{1}=(\alpha \cup \beta)$
$-R_{2}=(R-\beta)$


## Employee Information Example

$\square$ Combined schema:
employee(emp id, emp_name)
emp_info(emp_id, dependent, phone_num)
$\square$ Also have these dependencies:
■ emp_id $\rightarrow$ emp_name

- emp_id $\rightarrow$ dependent
- emp_id $\rightarrow$ phone_num
$\square$ emp_info is not in 4NF
$\square$ Following the rules for 4NF decomposition produces:
(emp_id, dependent)
(emp_id, phone_num)
$\square$ Note: Each relation's candidate key is the entire relation. The multivalued dependencies are trivial.


## Lossless Decompositions

$\square$ Can also define lossless decomposition with multivalued dependencies
$\square R_{1}$ and $R_{2}$ form a lossless decomposition of $R$ if at least one of these dependencies is in $D^{+}$:

$$
\begin{aligned}
& R_{1} \cap R_{2} \rightarrow R_{1} \\
& R_{1} \cap R_{2} \rightarrow R_{2}
\end{aligned}
$$

## Beyond Fourth Normal Form?

$\square$ Additional normal forms with various constraints
$\square$ Example: join dependencies
$\square$ Given $R$, and a decomposition $R_{1}$ and $R_{2}$ where $R_{1} \cup R_{2}=R:$
$\square$ The decomposition is lossless if, for all legal instances of $r(R)$, $\Pi_{R_{1}}(r) \bowtie \Pi_{R_{2}}(r)=r$
$\square$ Can state this as a join dependency: $*\left(R_{1}, R_{2}\right)$
$\square$ This is actually identical to a multivalued dependency!
$\square *\left(R_{1}, R_{2}\right)$ is equivalent to $R_{1} \cap R_{2} \rightarrow R_{1} \mid R_{2}$

## Join Dependencies and 5NF

$\square$ Join dependencies (JD) are a generalization of multivalued dependencies (MVD)
$\square$ Can specify JDs involving $N$ relation schemas, $N \geq 2$
$\square$ JDs are equivalent to MVDs when $N=2$
$\square$ Can easily construct JDs where $N>2$, with no equivalent set of MVDs
$\square$ Project-Join Normal Form (a.k.a. PJNF or 5NF):
$\square$ A relation schema $R$ is in PJNF with respect to a set of join dependencies $D$ if, for all JDs in $D^{+}$of the form $*\left(R_{1}, R_{2}, \ldots, R_{n}\right)$ where $R_{1} \cup R_{2} \cup \ldots \cup R_{n}=R$, at least one of the following holds:

- $*\left(R_{1}, R_{2}, \ldots, R_{n}\right)$ is a trivial join dependency
- Every $R_{i}$ is a superkey for $R$


## Join Dependencies and 5NF (2)

$\square$ If a schema is in Project-Join Normal Form then it is also in $4 N F$ (and thus, in BCNF)
$\square$ Every multivalued dependency is also a join dependency
$\square$ (Every functional dependency is also a multivalued dependency)
$\square$ One small problem:
$\square$ There isn't a complete, sound set of inference rules for join dependencies!
$\square$ Can't reason about our set of join dependencies D...
$\square$ This limits PJNF's real-world usefulness

## Domain-Key Normal Form

$\square$ Domain-key normal form (DKNF) is an even more general normal form, based on:
$\square$ Domain constraints: what values may be assigned to attribute $A$

- Usually inexpensive to test, even with CHECK constraints
$\square$ Key constraints: all attribute-sets $K$ that are a superkey for a schema $R$ (i.e. $K \rightarrow R$ )
- Almost always inexpensive to test
$\square$ General constraints: other predicates on valid relations in a schema
- Could be very expensive to test!
$\square$ A schema $R$ is in DKNF if the domain constraints and key constraints logically imply the general constraints
- An "ideal" normal form difficult to achieve in practice...

