

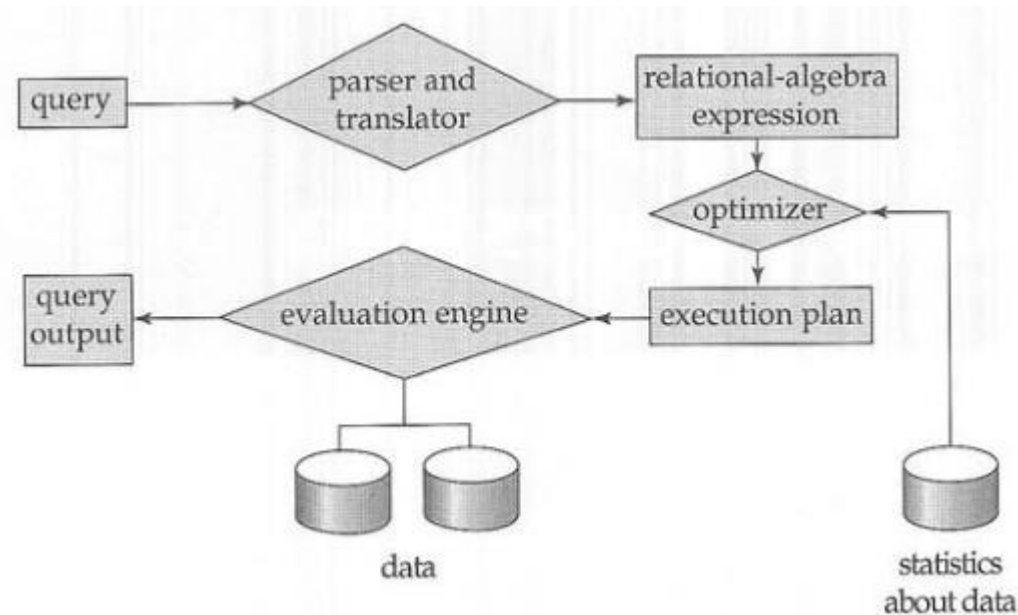
Query Equivalent

(Query processing Engine)

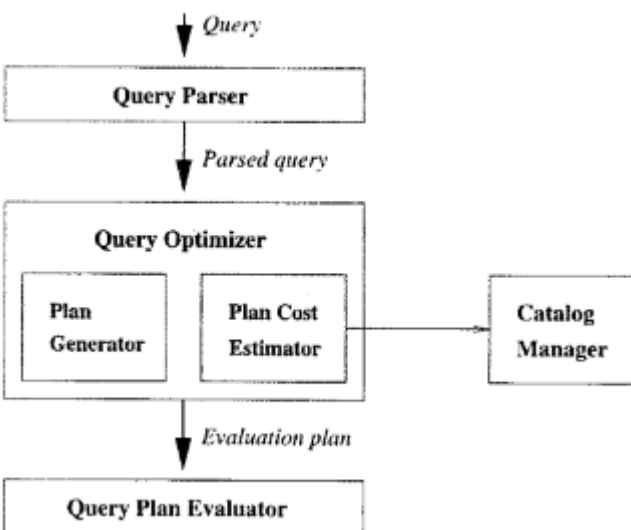
Adapted from

Chapter 14 – Database System Concept, Silberschatz, Korth, Sudarshan, 5th ed.

Chapter 15 – Database Management System, Ramakrishnan, Gehrke, 3rd ed



Query Evaluation Engine



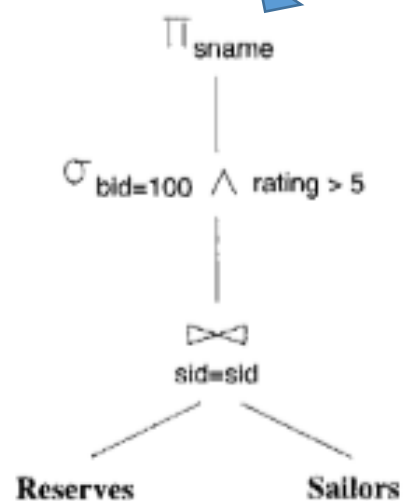
Sailors(sid: integer, sname: string, rating: integer, age: real) Database
Reserves(sid: integer, bid: integer, day: dates, rname: string)

Query

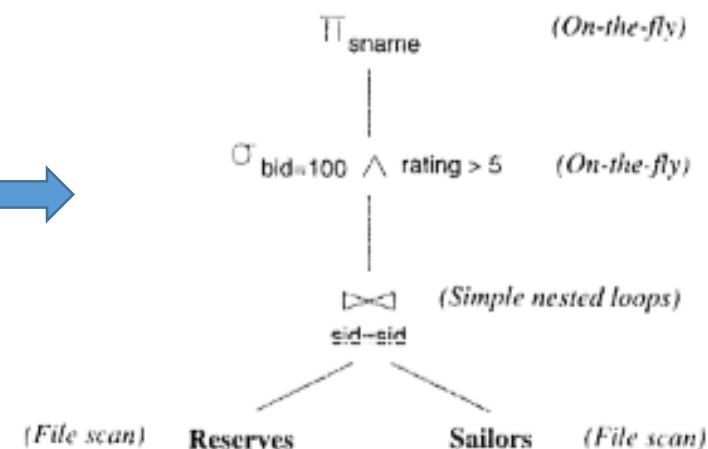
```
SELECT S.sname
FROM   Reserves R, Sailors S
WHERE  R.sid = S.sid
       AND R.bid = 100 AND S.rating > 5
```

$\pi_{sname}(\sigma_{bid=100 \wedge rating > 5}(Reserves \bowtie_{sid=sid} Sailors))$

Relational Algebra

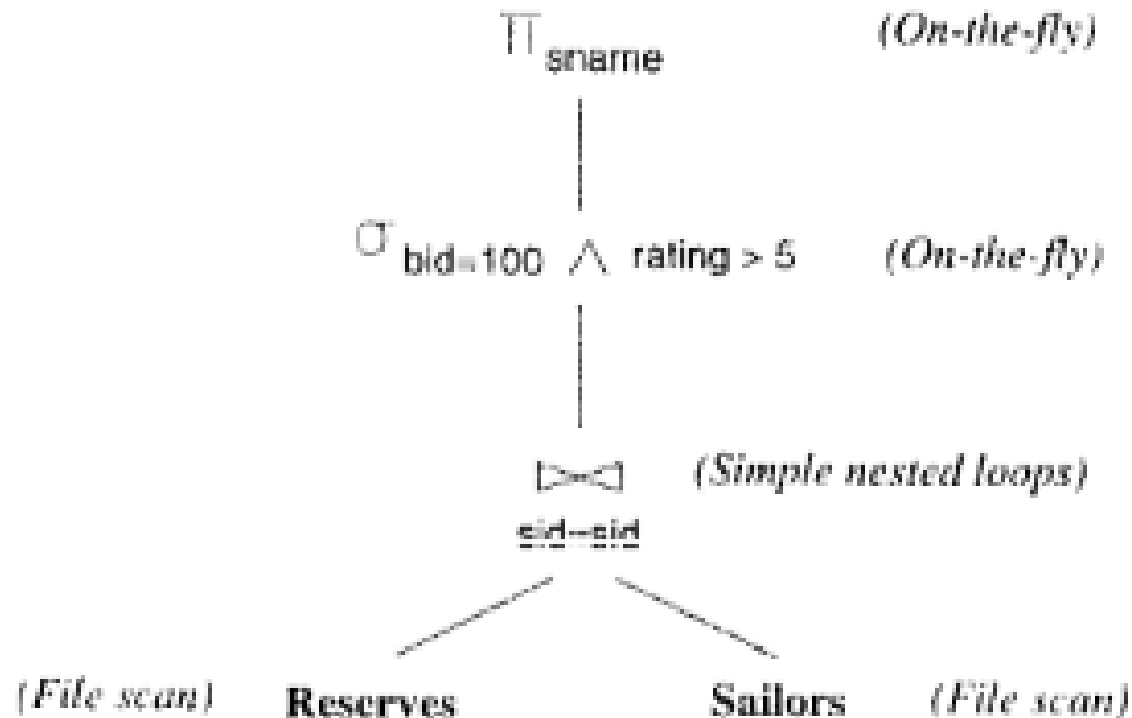


Relational Algebra tree



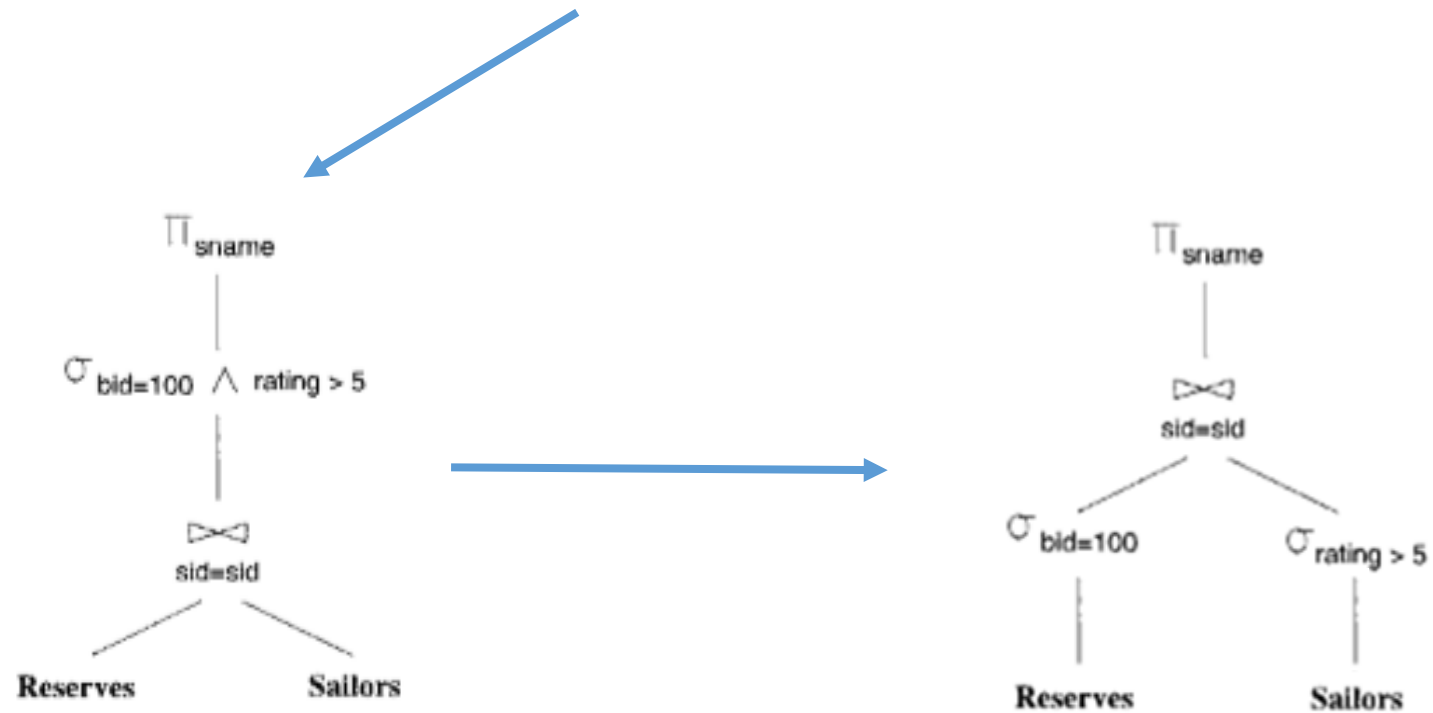
Query Evaluation Plan

An **evaluation plan** defines exactly what algorithm is used for each operation, and how the execution of the operations is coordinated.



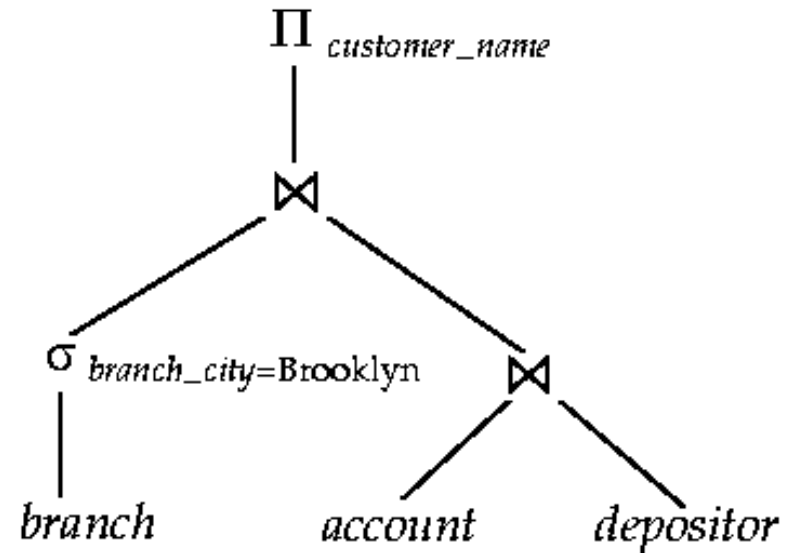
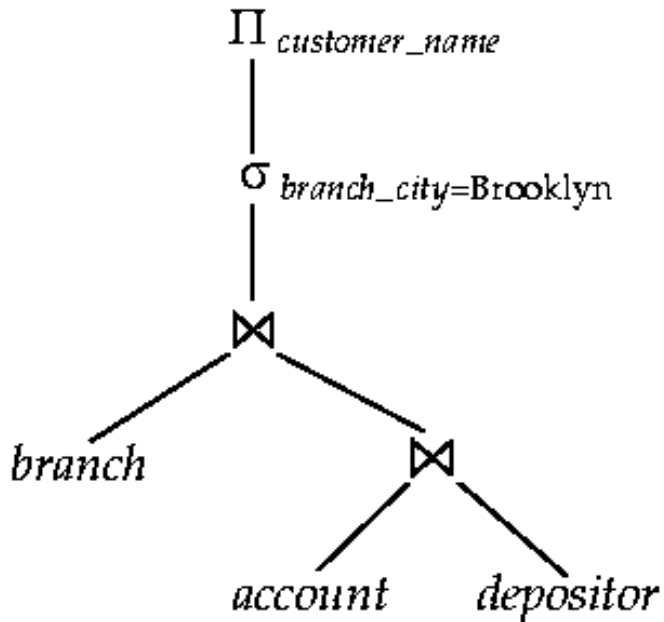
A query can be served with multiple transformations (**execution rules**) providing the same results

$$\pi_{sname}(\sigma_{bid=100 \wedge rating > 5}(Reserves \bowtie_{sid=sid} Sailors))$$



A query can be served with various **execution plans** providing the same results

$\Pi_{customer_name} (\sigma_{branch_city = \text{"Brooklyn"}} (branch \bowtie (account \bowtie depositor)))$



- Cost difference between evaluation plans for a query can be enormous
 - E.g. seconds vs. days in some cases
- Steps in **cost-based query optimization**
 1. Generate logically equivalent expressions using **equivalence rules**
 2. Annotate resultant expressions to get alternative query plans
 3. Choose the cheapest plan based on **estimated cost**

Transformation of Relational Expressions

- Two relational algebra expressions are said to be **equivalent** if the two expressions generate the same set of tuples on every legal database instance
 - order of tuples is irrelevant
- In SQL, inputs and outputs are **multisets** of tuples
 - Two expressions in the multiset version of the relational algebra are said to be equivalent if the two expressions generate the same multiset of tuples on every legal database instance.
- An **equivalence rule** says that expressions of two forms are equivalent
 - Can replace expression of first form by second, or vice versa

Equivalence Rules

1. Conjunctive selection operations can be deconstructed into a sequence of individual selections (**cascading of selection**).

$$\sigma_{c_1 \wedge c_2 \wedge \dots \wedge c_n}(R) \equiv \sigma_{c_1}(\sigma_{c_2}(\dots(\sigma_{c_n}(R))\dots))$$

Equivalence Rules

1. Conjunctive selection operations can be deconstructed into a sequence of individual selections (**cascading of selection**).

$$\sigma_{\theta_1 \wedge \theta_2}(R) = \sigma_{\theta_1}(\sigma_{\theta_2}(R))$$

EiD	Name	Salary	Age
01	Ram	30000	35
02	Shyam	10000	25
03	Kumar	25000	30
04	Ram	50000	45
05	John	75000	50
05	Jack	5000	20



$\sigma_{\text{Salary} > 20000 \text{ AND Age} < 40}(R)$

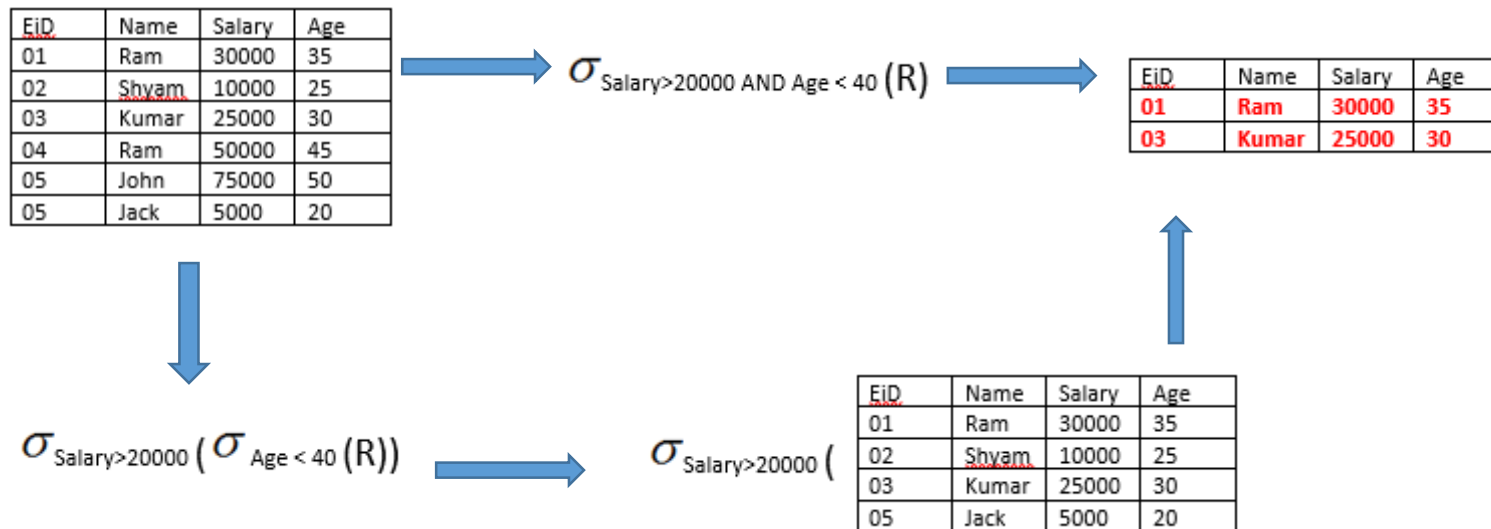


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Equivalence Rules

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Equivalence Rules

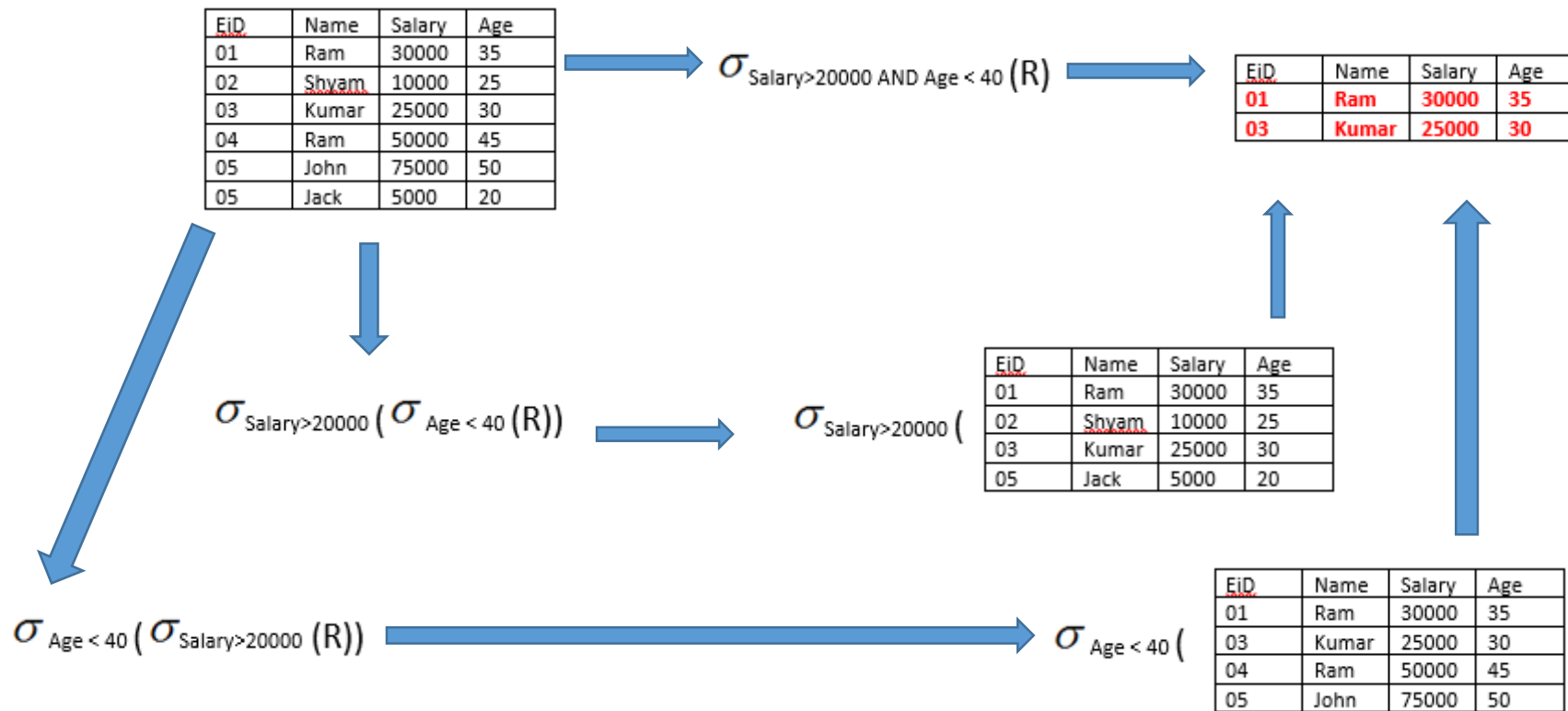
2. Selection operations are **commutative**.

$$\sigma_{\theta_1}(\sigma_{\theta_2}(R)) = \sigma_{\theta_2}(\sigma_{\theta_1}(R))$$

Equivalence Rules

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$$\sigma_{\theta_1}(\sigma_{\theta_2}(R)) = \sigma_{\theta_2}(\sigma_{\theta_1}(R))$$



Equivalence Rules

3. Only the last in a sequence of projection operations is needed, the others can be omitted (**cascading of projection**).

$$\Pi_{L_1}(\Pi_{L_2}(\dots(\Pi_{L_n}(E))\dots)) = \Pi_{L_1}(E)$$

Equivalence Rules

4. Selections can be combined with Cartesian products and theta joins.

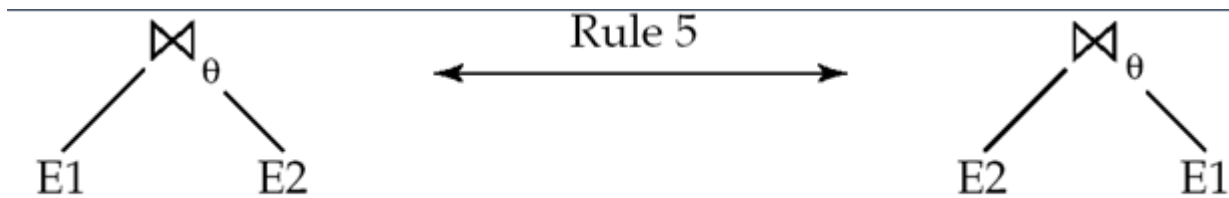
a. $\sigma_{\theta}(E_1 \times E_2) = E_1 \bowtie_{\theta} E_2$

b. $\sigma_{\theta_1}(E_1 \bowtie_{\theta_2} E_2) = E_1 \bowtie_{\theta_1 \wedge \theta_2} E_2$

Equivalence Rules

5. Theta-join operations (and natural joins) are **commutative**.

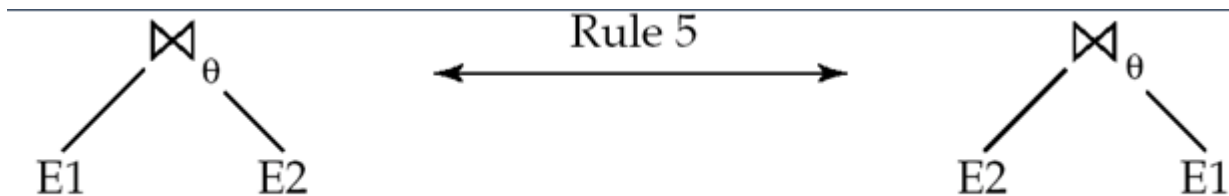
$$E_1 \bowtie_{\theta} E_2 = E_2 \bowtie_{\theta} E_1$$



Equivalence Rules

5. Theta-join operations (and natural joins) are **commutative**.

$$E_1 \bowtie_{\theta} E_2 = E_2 \bowtie_{\theta} E_1$$



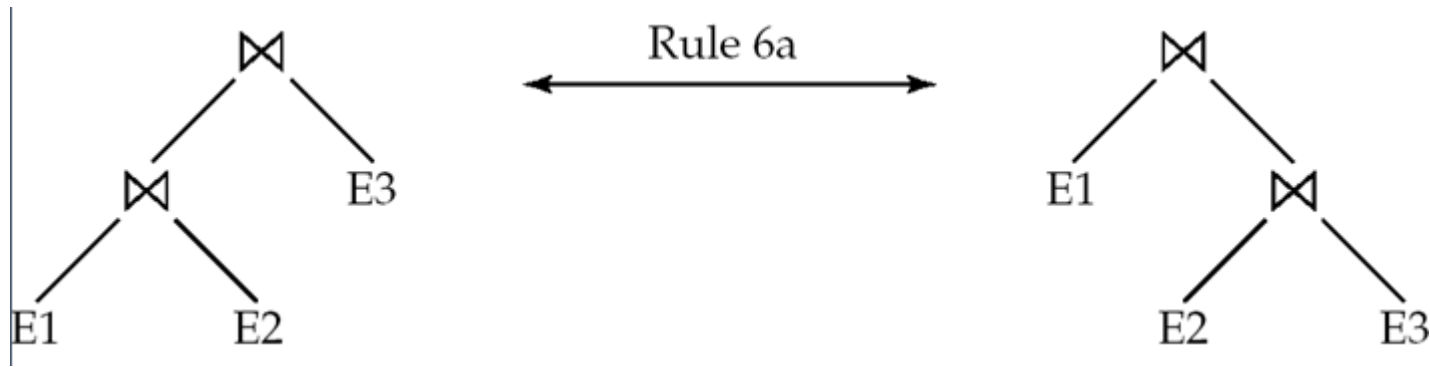
Cartesian Product also holds the following

$$R \times S \equiv S \times R$$

Equivalence Rules (Cont.)

6. (a) Natural join operations are **associative**:

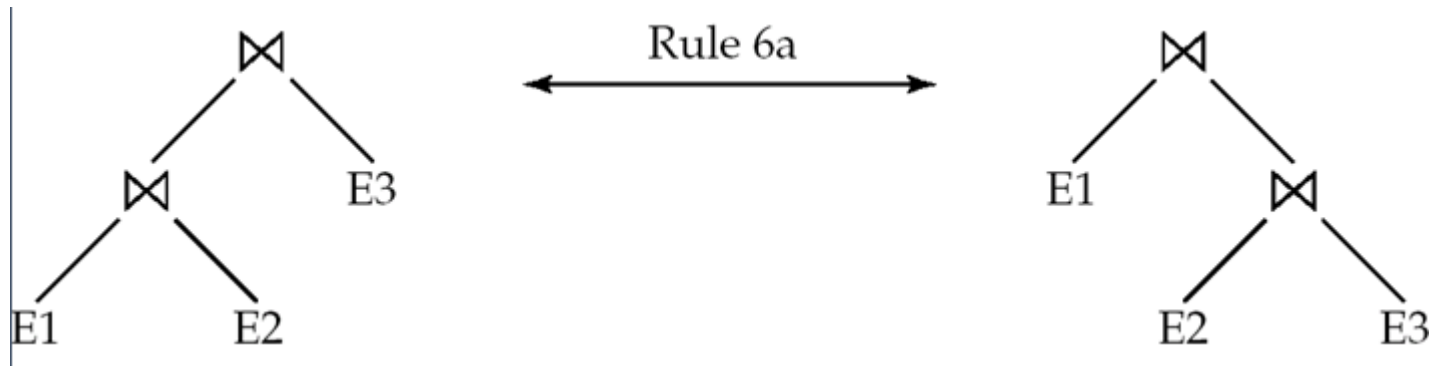
$$(E_1 \bowtie E_2) \bowtie E_3 = E_1 \bowtie (E_2 \bowtie E_3)$$



Equivalence Rules (Cont.)

6. (a) Natural join operations are **associative**:

$$(E_1 \bowtie E_2) \bowtie E_3 = E_1 \bowtie (E_2 \bowtie E_3)$$



Natural Join also holds the following

$$R \bowtie (S \bowtie T) \equiv (T \bowtie R) \bowtie S$$

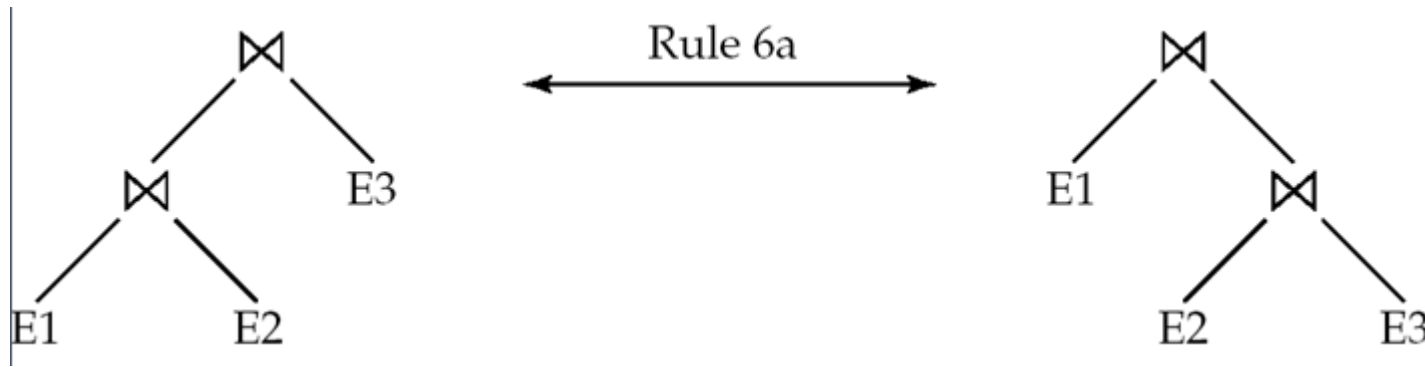
Equivalence Rules (Cont.)

6. (a) Natural join operations are **associative**:

$$(E_1 \bowtie E_2) \bowtie E_3 = E_1 \bowtie (E_2 \bowtie E_3)$$

Cartesian Product also holds the following

$$R \times (S \times T) \equiv (R \times S) \times T$$



Natural Join also holds the following

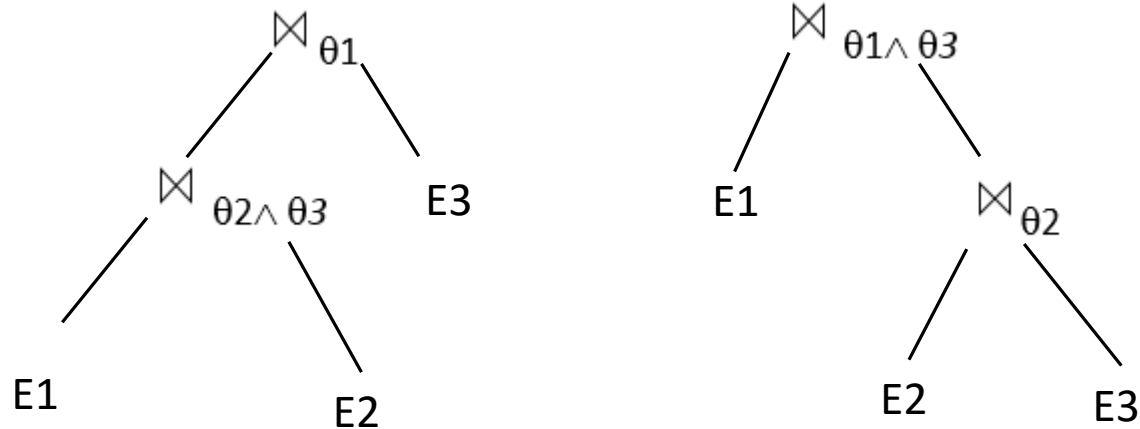
$$R \bowtie (S \bowtie T) \equiv (T \bowtie R) \bowtie S$$

Equivalence Rules (Cont.)

6.(b) Theta joins are associative in the following manner:

$$(E_1 \bowtie_{\theta_1} E_2) \bowtie_{\theta_2 \wedge \theta_3} E_3 = E_1 \bowtie_{\theta_1 \wedge \theta_3} (E_2 \bowtie_{\theta_2} E_3)$$

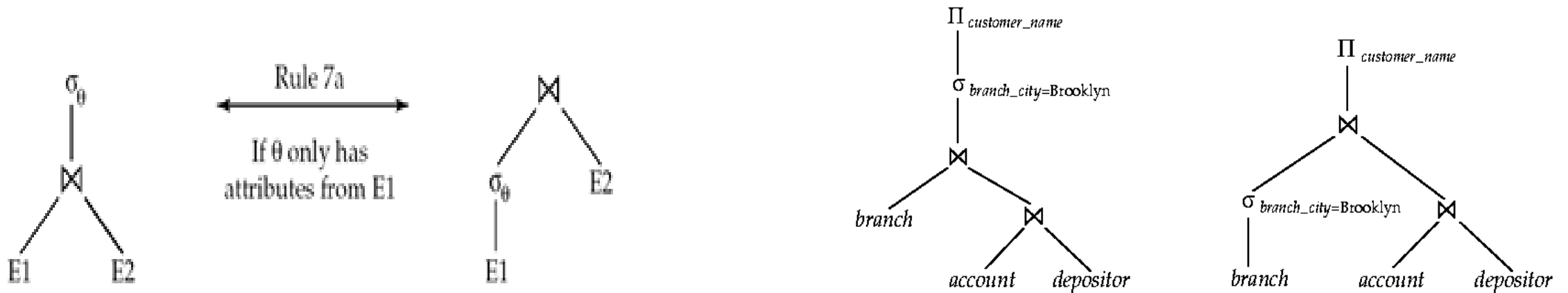
where θ_2 involves attributes from only E_2 and E_3 .



Equivalence Rules (Cont.)

7. The selection operation distributes over the theta join operation under the following two conditions:
- (a) When all the attributes in θ_0 involve only the attributes of one of the expressions (E_1) being joined.

$$\sigma_{\theta_0}(E_1 \bowtie_{\theta} E_2) = (\sigma_{\theta_0}(E_1)) \bowtie_{\theta} E_2$$

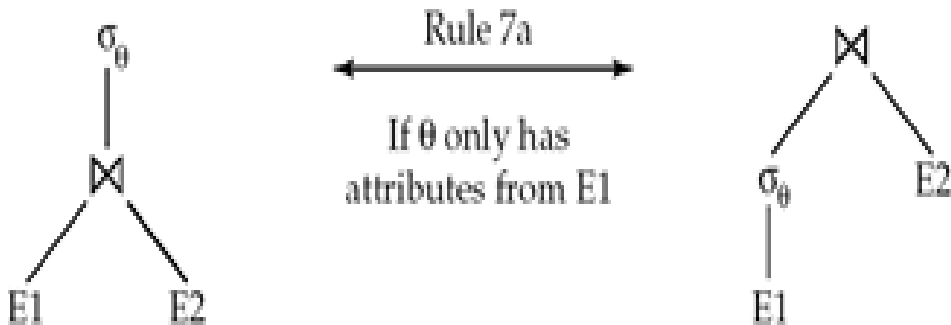


Equivalence Rules (Cont.)

7. The selection operation distributes over the theta join operation under the following two conditions:
- (a) When all the attributes in θ_0 involve only the attributes of one of the expressions (E_1) being joined.

$$\sigma_{\theta_0}(E_1 \bowtie_{\theta} E_2) = (\sigma_{\theta_0}(E_1)) \bowtie_{\theta} E_2$$

Also true for Cartesian Product

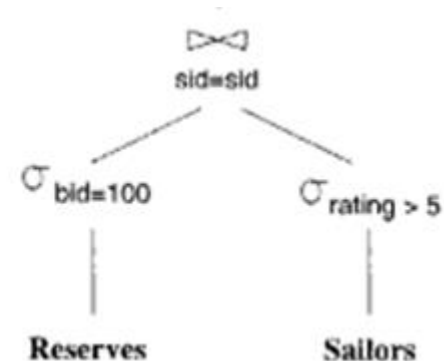
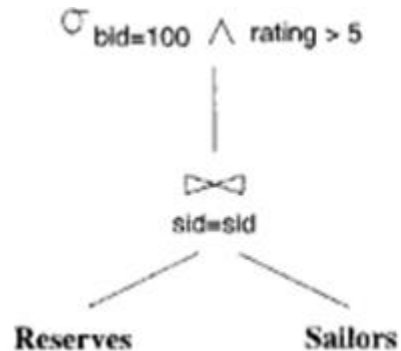


Equivalence Rules (Cont.)

7. The selection operation distributes over the theta join operation under the following two conditions:

(b) When θ_1 involves only the attributes of E_1 and θ_2 involves only the attributes of E_2 .

$$\sigma_{\theta_1 \wedge \theta_2} (E_1 \bowtie_{\theta} E_2) = (\sigma_{\theta_1}(E_1)) \bowtie_{\theta} (\sigma_{\theta_2}(E_2))$$



Projection over Selection

Can we perform?

$$\pi_a(\sigma_c(R)) \equiv \sigma_c(\pi_a(R))$$

Projection over Selection

Can we perform?

$$\pi_a(\sigma_c(R)) \equiv \sigma_c(\pi_a(R))$$

Yes, if **c** is included in **a**

Projection over Selection

Can we perform?

$$\Pi_{\theta_1, \theta_2}(\mathbf{E}_1 \bowtie_{\theta} \mathbf{E}_2) = (\Pi_{\theta_1}(\mathbf{E}_1)) \bowtie_{\theta} (\Pi_{\theta_2}(\mathbf{E}_2))$$

Projection over Selection

Can we perform?

$$\Pi_{\theta_1, \theta_2} (E_1 \bowtie_{\theta} E_2) = (\Pi_{\theta_1}(E_1)) \bowtie_{\theta} (\Pi_{\theta_2}(E_2))$$

$$\Pi_{\theta_1, \theta_2} (E_1 \bowtie_{\theta} E_2) = (\Pi_{\theta_1, \theta_2}(E_1)) \bowtie_{\theta} (\Pi_{\theta_1, \theta_2}(E_2))$$

8. The projection operation distributes over the theta join operation as follows:

(a) if θ involves only attributes from $L_1 \cup L_2$:

$$\Pi_{L_1 \cup L_2} (E_1 \bowtie_{\theta} E_2) = (\Pi_{L_1} (E_1)) \bowtie_{\theta} (\Pi_{L_2} (E_2))$$

(b) Consider a join $E_1 \bowtie_{\theta} E_2$.

- Let L_1 and L_2 be sets of attributes from E_1 and E_2 , respectively.
- Let L_3 be attributes of E_1 that are involved in join condition θ , but are not in $L_1 \cup L_2$, and
- let L_4 be attributes of E_2 that are involved in join condition θ , but are not in $L_1 \cup L_2$.

$$\Pi_{L_1 \cup L_2} (E_1 \bowtie_{\theta} E_2) = \Pi_{L_1 \cup L_2} ((\Pi_{L_1 \cup L_3} (E_1)) \bowtie_{\theta} (\Pi_{L_2 \cup L_4} (E_2)))$$

9. The set operations union and intersection are **commutative**

$$E_1 \cup E_2 = E_2 \cup E_1$$

$$E_1 \cap E_2 = E_2 \cap E_1$$

10. Set union and intersection are **associative**.

$$(E_1 \cup E_2) \cup E_3 = E_1 \cup (E_2 \cup E_3)$$

$$(E_1 \cap E_2) \cap E_3 = E_1 \cap (E_2 \cap E_3)$$

11. The selection operation distributes over \cup , \cap and $-$.

$$\sigma_{\theta}(E_1 - E_2) = \sigma_{\theta}(E_1) - \sigma_{\theta}(E_2)$$

and similarly for \cup and \cap in place of $-$

Also: $\sigma_{\theta}(E_1 - E_2) = \sigma_{\theta}(E_1) - E_2$

and similarly for \cap in place of $-$, but not for \cup

12. The projection operation distributes over union

$$\Pi_L(E_1 \cup E_2) = (\Pi_L(E_1)) \cup (\Pi_L(E_2))$$

Example:

branch(branch_name, branch_city, assets),
account(account_number, branch_name, balance)
depositor(customer_name, account_number)

- Query: Find the names of all customers who have an account at some branch located in Brooklyn.

$$\Pi_{customer_name}(\sigma_{branch_city = \text{"Brooklyn"}}(branch \bowtie (account \bowtie depositor)))$$

- Transformation using rule 7a.

$$\Pi_{customer_name}((\sigma_{branch_city = \text{"Brooklyn"}}(branch)) \bowtie (account \bowtie depositor))$$

- Can also be transformed as

$$\sigma_{branch_city = \text{"Brooklyn"}}(\Pi_{customer_name, branch_city}(branch \bowtie (account \bowtie depositor)))$$

Example with Multiple Transformations

- Query: Find the names of all customers with an account at a Brooklyn branch whose account balance is over \$1000.

$$\Pi_{customer_name}(\sigma_{branch_city = "Brooklyn" \wedge balance > 1000} (branch \bowtie (account \bowtie depositor)))$$

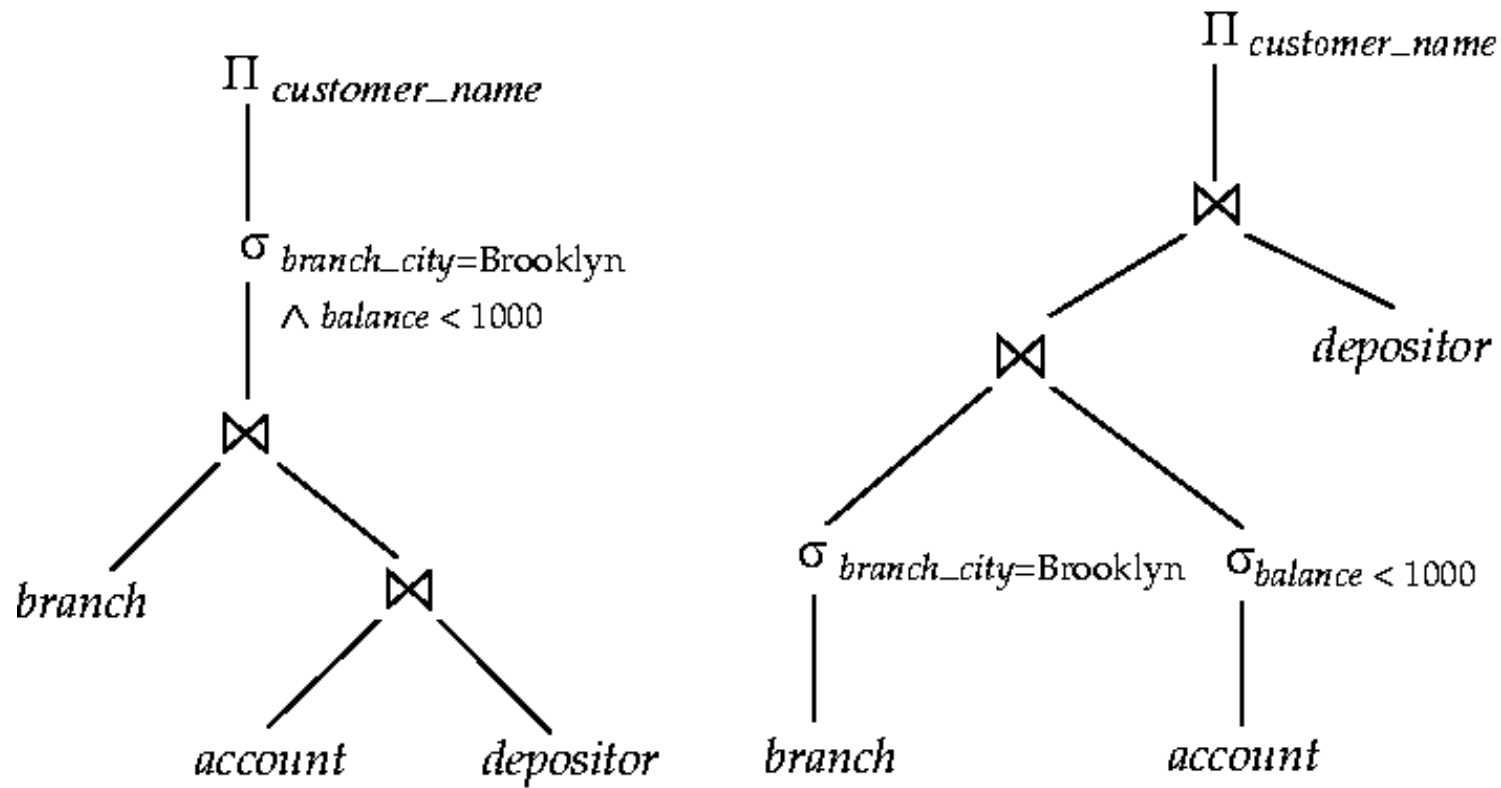
- Transformation using join associatively (Rule 6a):

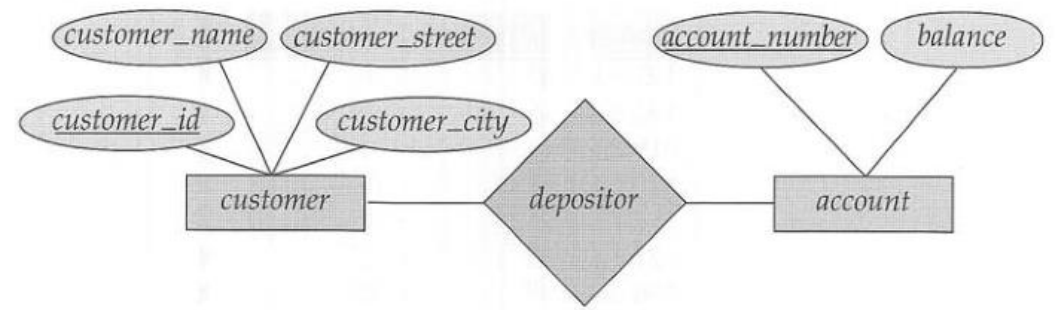
$$\Pi_{customer_name}((\sigma_{branch_city = "Brooklyn" \wedge balance > 1000} (branch \bowtie account)) \bowtie depositor)$$

- Second form provides an opportunity to apply the “perform selections early” rule, resulting in the subexpression

$$\sigma_{branch_city = "Brooklyn"} (branch) \bowtie \sigma_{balance > 1000} (account)$$

Multiple Transformations (Cont.)





$$\Pi_{customer_name}(\sigma_{branch_city = \text{"Brooklyn"}}(branch \bowtie (account \bowtie depositor)))$$

$$\Pi_{customer_name}((\sigma_{branch_city = \text{"Brooklyn"}}(branch) \bowtie account) \bowtie depositor)$$

$$\Pi_{customer_name}((\Pi_{account_number}((\sigma_{branch_city = \text{"Brooklyn"}}(branch) \bowtie account) \bowtie depositor))$$