

Chapter 7(contd..): Query Optimization

Database System Concepts 5th Ed.

©Silberschatz, Korth and Sudarshan See www.db-book.com for conditions on re-use





Chapter 14: Query Optimization

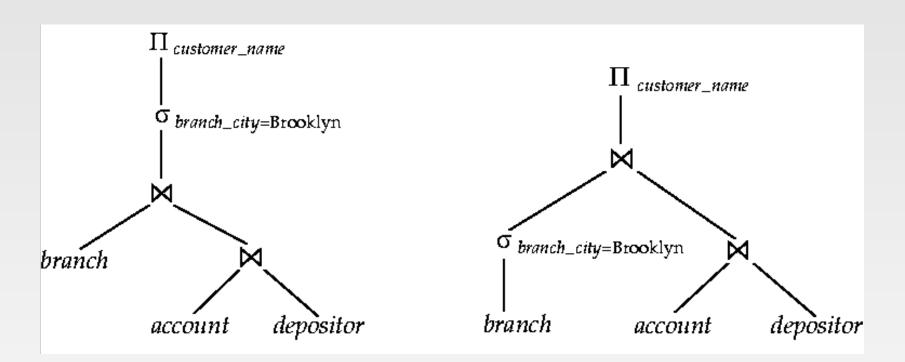
- Introduction
- Transformation of Relational Expressions





Introduction

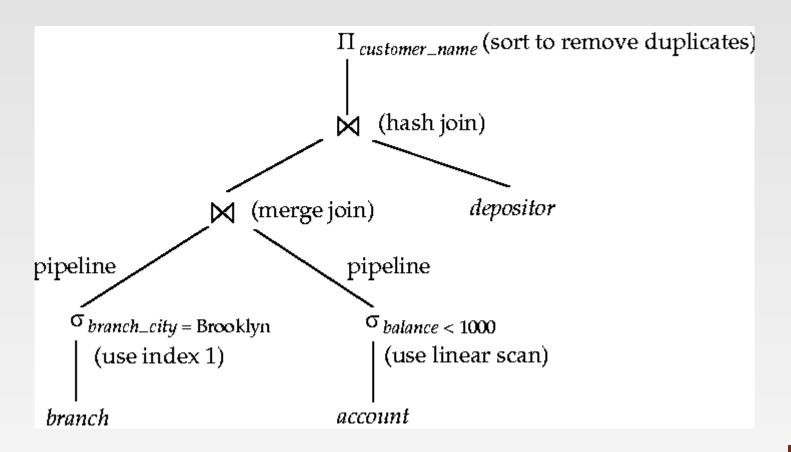
- Alternative ways of evaluating a given query
 - Equivalent expressions
 - Different algorithms for each operation





Introduction (Cont.)

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





Introduction (Cont.)

- 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**
- Estimation of plan cost based on:
 - Statistical information about relations. Examples:
 - number of tuples, number of distinct values for an attribute
 - Statistics estimation for intermediate results
 - to compute cost of complex expressions
 - Cost formulae for algorithms, computed using statistics





Generating Equivalent Expressions

Database System Concepts 5th Ed.

©Silberschatz, Korth and Sudarshan See www.db-book.com for conditions on re-use





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
 - Note: 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.

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

2. Selection operations are commutative.

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

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

$$\Pi_{L_1}(\Pi_{L_2}(...(\Pi_{Ln}(E))...))=\Pi_{L_1}(E)$$

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

$$\alpha$$
. $\sigma_{\theta}(E_1X E_2) = E_1 \bowtie_{\theta} E_2$

β.
$$\sigma_{\theta_1}(E_1 \bowtie_{\theta_2} E_2) = E_1 \bowtie_{\theta_1 \land \theta_2} E_2$$



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

$$E_1^{\downarrow} \cup_{\theta} E_2 = E_2 \quad || \cup_{\theta} E_1|$$

6. (a) Natural join operations are associative:

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

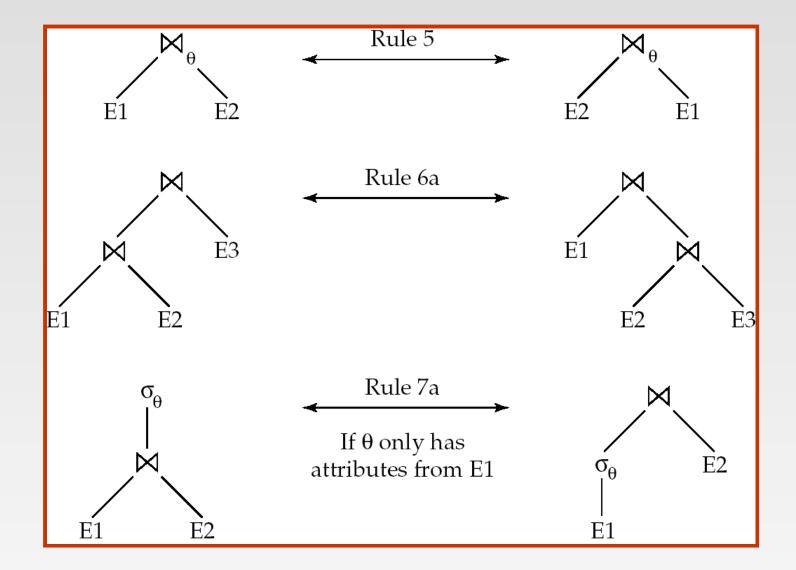
(b) Theta joins are associative in the following manner:

$$\triangleright \langle E_1 | E_2 \rangle$$
 $= E_1 \mid E_3 \rangle$ $= E_1 \mid E_3 \rangle$

where θ_2 involves attributes from only E_2 and E_3 .



Pictorial Depiction of Equivalence Rules





- 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.

$$\sigma_{\theta}(\mathsf{E}_1^{\bowtie}_{\theta} \mathsf{E}_2) = (\sigma_{\theta}(\mathsf{E}_1))^{\bowtie}_{\theta} \mathsf{E}_2$$

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

only the attributes of
$$E_2$$
.

$$\sigma_{\theta_1 \uparrow \theta_2}(E_1 \quad \theta_1 E_2) = (\sigma_{\theta_1}(E_1)) \quad (\sigma_{\theta_2}(E_2))$$



- 8. The projection operation distributes over the theta join operation as follows:
 - (a) if θ involves only attributes from $L_1 \cup L_2$:

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

- (b) Consider a join $E_1 \bowtie E_2$.
 - 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 $L_1 \cup L_2$.

$$\Pi_{L_{1}\cup L_{2}}^{1}(E_{1}\bowtie_{\theta}E_{2})=\Pi_{L_{1}\cup L_{2}}((\Pi_{L_{1}\cup L_{3}}(E_{1}))\bowtie_{\theta}(\Pi_{L_{2}\cup L_{4}}(E_{2})))$$





1. The set operations union and intersection are commutative

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

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

- (set difference is not commutative).
- 2. Set union and intersection are associative.

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

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

1. 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

$$\Pi_{\perp}(E_1 \cup E_2) = (\Pi_{\perp}(E_1)) \cup (\Pi_{\perp}(E_2))$$





Transformation Example: Pushing Selections

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

```
\Pi_{\text{customer\_name}}(\sigma_{\text{branch\_city = "Brooklyn"}})

(branch \bowtie (account \bowtie depositor)))
```

Transformation using rule 7a.

```
 \begin{array}{ccc} \Pi_{\text{customer\_name}} \\ & ((\sigma_{\text{branch\_city ="Brooklyn"}}(\textit{branch})) \\ & (account & depositor)) \end{array}
```

Performing the selection as early as possible reduces the size of the relation to be joined.



Example with Multiple Transformations

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

$$\Pi_{customer_name}(\sigma_{branch_city = "Brooklyn" \land balance > 1000})$$

(branch (account depositor)))

Transformation using join associatively (Rule 6a):

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

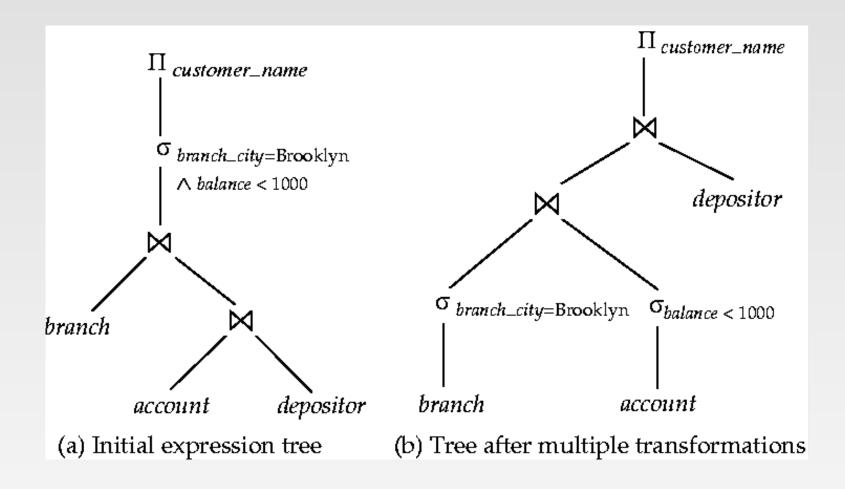
Second form provides an opportunity to apply the "perform selections early" rule, resulting in the subexpression

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

Thus a sequence of transformations can be useful



Multiple Transformations (Cont.)







Transformation Example: Pushing Projections

$$\Pi_{customer_name}((\sigma_{branch_city = "Brooklyn"}(branch)) \bowtie (count) \bowtie (count)$$

When we compute

$$(\sigma_{branch \ city = "Brooklyn"} (branch) \ account)$$

we obtain a relation whose schema is: (branch_name, branch_city, assets, account_number, balance)

Push projections using equivalence rules 8a and 8b; eliminate unneeded attributes from intermediate results to get:

```
\Pi_{\text{customer\_name}}((\sigma_{\text{branch\_city = "Brooklyn"}}(branch) \quad account))
```

Performing the projection as early as possible reduces the size of the relation to be joined.





End of Chapter

Database System Concepts 5th Ed.

©Silberschatz, Korth and Sudarshan See www.db-book.com for conditions on re-use

