## 4. Logical optimization

## Logical optimization

- Algebraic Laws
- Improving the logical query plan


## Algebraic laws

- Commutative and associative laws
- Laws involving selection
- Laws involving projection
- Laws about joins and product


## Commutative and Associative Laws

| Operation | Commutativity | Associativity |
| :--- | :---: | :---: |
| Cartesian product | $R \times S=S \times R^{1)}$ | $(R \times S) \times T=R \times(S \times T)$ |
| Natural join | $R \bowtie S=S \bowtie R^{1)}$ | $(R \bowtie S) \bowtie T=R \bowtie(S \bowtie T)^{2)}$ |
| Union | $R \cup S=S \cup R$ | $(R \cup S) \cup T=R \cup(S \cup T)$ |
| Intersection | $R \cap S=S \cap R$ | $(R \cap S) \cap T=R \cap(S \cap T)$ |

${ }^{1)}$ The order of columns changes.
${ }^{2)}$ Natural join of three relations is performed on attributes which are common for all the relations.

## Restricted associativity of theta-join

$$
\begin{array}{|c|c}
\hline \text { Commutativity } & \text { Associativity } \\
\hline R \underset{\theta}{\bowtie} S=\underset{\theta}{\bowtie} R & ? \\
\hline
\end{array}
$$

$$
R(A, B) ; S(C, D) ; T(E, F)
$$

$$
\begin{aligned}
& R \underset{B=C}{\bowtie}(S \underset{D=E}{\bowtie} T)=(R \underset{B=C}{\bowtie} S) \underset{D=E}{\bowtie} T \\
& R \underset{A=F}{\bowtie}(S \underset{D=E}{\bowtie} T) \neq(R \underset{A=E}{\bowtie} S) \underset{D=E}{\bowtie} T
\end{aligned}
$$

## Laws involving selection

1. $\sigma_{C_{1} \& C_{2}}(R)=\sigma_{C_{1}}\left(\sigma_{C_{2}}(R)\right)$
2. $\sigma_{C_{1}}\left(\sigma_{C_{2}}(R)\right)=\sigma_{C_{2}}\left(\sigma_{C_{1}}(R)\right)$
3. $\sigma_{A<x}(R \underset{(A)}{\ltimes} S)=\left(\sigma_{A<x}(R)\right) \underset{(A)}{\bowtie}\left(\sigma_{A<x}(S)\right)$

## Laws involving projection

1. $\pi_{\alpha}(R \bowtie S)=\pi_{\alpha}\left(\pi_{\beta}(R) \bowtie \pi_{\gamma}(S)\right)$
2. $\pi_{\alpha}(\underset{\theta}{\text { ® }} S)=\pi_{\alpha}\left(\pi_{\beta}(R) \underset{\theta}{\bowtie} \pi_{\gamma}(S)\right)$
3. $\pi_{\alpha}(R \times S)=\pi_{\alpha}\left(\pi_{\beta}(R) \times \pi_{\gamma}(S)\right)$

## Projection over natural join

$$
\pi_{\alpha}(R \bowtie S)=\pi_{\alpha}\left(\pi_{\beta}(R) \bowtie \pi_{\gamma}(S)\right)
$$

$\beta$ - the join attributes and the attributes of $\alpha$ that are found among the attributes of $R$
$\gamma$ - the join attributes and the attributes of $\alpha$ that are found among the attributes of $S$

## Projection over theta-join

$$
\pi_{\alpha}(R \underset{\theta}{\bowtie} S)=\pi_{\alpha}\left(\pi_{\beta}(R) \bowtie \pi_{\gamma}(S)\right)
$$

$\beta$ - the join attributes (i.e., those mentioned in condition $\theta$ ) and the attributes of $\alpha$ that are found among the attributes of $R$
$\gamma$ - the join attributes (i.e., those mentioned in condition $\theta$ ) and the attributes of $\alpha$ that are found among the attributes of $S$

## Projection over cartesian product

$$
\pi_{\alpha}(R \times S)=\pi_{\alpha}\left(\pi_{\beta}(R) \times \pi_{\gamma}(S)\right)
$$

$\beta$ - the attributes of $\alpha$ that are found among the attributes of $R$
$\gamma$ - the attributes of $\alpha$ that are found among the attributes of $S$

## Laws about joins and product

$$
\begin{aligned}
& \text { 1. } \quad R \underset{\theta}{\bowtie} S=\sigma_{\theta}(R \times S) \\
& \text { 2. } \quad R \underset{(A)}{\bowtie} S=\pi_{R . *, S . *-S . A}\left(\sigma_{R . A=S . A}(R \times S)\right)
\end{aligned}
$$

$$
R \underset{\mathrm{C}>\mathrm{E}^{*} 1000}{\bowtie} S=\sigma_{\mathrm{C}>\mathrm{E}^{*} 1000}(R \times S)
$$

| R |  |  | S |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathrm{A}^{*}$ | B | C | D* | F | E |
| 1 | 20 | 600 | 1 | 3 | 0.2 |
| 2 | 40 | 300 | 2 | 1 | 0.7 |
| 3 | 20 | 150 | 3 | 1 | 0.5 |
| 4 | 10 | 300 |  |  |  |


| $\underset{\substack{\mathrm{CP} \mathbb{F}^{1000} \\ \mathrm{~S}}}{ }$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathrm{A}^{*}$ | B | C | D* | F | E |
| 1 | 20 | 600 | 1 | 3 | 0.2 |
| 1 | 20 | 600 | 3 | 1 | 0.5 |
| 2 | 40 | 300 | 1 | 3 | 0.2 |
| 4 | 10 | 300 | 1 | 3 | 0.2 |


| 1) $Q=R \times S$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{A}^{*}$ | $\mathbf{B}$ | $\mathbf{C}$ | $\mathbf{D}^{*}$ | $\mathbf{F}$ | $\mathbf{E}$ |
| 1 | 20 | 600 | 1 | 3 | 0.2 |
| 1 | 20 | 600 | 2 | 1 | 0.7 |
| 1 | 20 | 600 | 3 | 1 | 0.5 |
| 2 | 40 | 300 | 1 | 3 | 0.2 |
| 2 | 40 | 300 | 2 | 1 | 0.7 |
| 2 | 40 | 300 | 3 | 1 | 0.5 |
| 3 | 20 | 150 | 1 | 3 | 0.2 |
| 3 | 20 | 150 | 2 | 1 | 0.7 |
| 3 | 20 | 150 | 3 | 1 | 0.5 |
| 4 | 10 | 300 | 1 | 3 | 0.2 |
| 4 | 10 | 300 | 2 | 1 | 0.7 |
| 4 | 10 | 300 | 3 | 1 | 0.5 |

2) $\sigma_{\mathrm{C}>\mathrm{E} * 1000}(Q)$

| $\mathbf{A}^{*}$ | $\mathbf{B}$ | $\mathbf{C}$ | $\mathbf{D}^{*}$ | $\mathbf{F}$ | $\mathbf{E}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 20 | 600 | 1 | 3 | 0.2 |
| 1 | 20 | 600 | 3 | 1 | 0.5 |
| 2 | 40 | 300 | 1 | 3 | 0.2 |
| 4 | 10 | 300 | 1 | 3 | 0.2 |

$$
R \underset{(A)}{\bowtie} S=\pi_{R . *, S . *-S . A}\left(\sigma_{R . A=S . A}(R \times S)\right)
$$

| R |  |  |  | S |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathrm{A}^{*}$ | B | C |  | D* | $\mathrm{A}^{\text {\# }}$ | E |
| 1 | 20 | 100 |  | 1 | 3 | 0.2 |
| 2 | 40 | 300 | x | 2 | 1 | 0.5 |
| 3 | 20 | 100 |  | 3 | 1 | 0.5 |
| 4 | 10 | 300 | $\mathrm{W}=\mathrm{R} \bowtie$ ( $\mathrm{S}^{\text {d }}$ |  |  |  |
|  |  |  | B | C | D* | E |
|  |  | 1 | 20 | 100 | 2 | 0.5 |
|  |  | 1 | 20 | 100 | 3 | 0.5 |
|  |  | 3 | 20 | 100 | 1 | 0.2 |


| 1) $Q=R \times S$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| R.A* | B | C | D* | S. ${ }^{\text {\# }}$ | E |
| 1 | 20 | 100 | 1 | 3 | 0.2 |
| 1 | 20 | 100 | 2 | 1 | 0.5 |
| 1 | 20 | 100 | 3 | 1 | 0.5 |
| 2 | 40 | 300 | 1 | 3 | 0.2 |
| 2 | 40 | 300 | 2 | 1 | 0.5 |
| 2 | 40 | 300 | 3 | 1 | 0.5 |
| 3 | 20 | 100 | 1 | 3 | 0.2 |
| 3 | 20 | 100 | 2 | 1 | 0.5 |
| 3 | 20 | 100 | 3 | 1 | 0.5 |
| 4 | 10 | 300 | 1 | 3 | 0.2 |
| 4 | 10 | 300 | 2 | 1 | 0.5 |
| 4 | 10 | 300 | 3 | 1 | 0.5 |

3) $\mathrm{W}=\pi_{R, *, S, *-S . A}(P)$

| $\mathbf{A}^{*}$ | $\mathbf{B}$ | $\mathbf{C}$ | $\mathbf{D}^{*}$ | $\mathbf{E}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1 | 20 | 100 | 2 | 0.5 |
| 1 | 20 | 100 | 3 | 0.5 |
| 3 | 20 | 100 | 1 | 0.2 |

2) $P=\sigma_{R . A=S . A}(Q)$

|  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{R}^{*} \mathbf{A}^{*}$ | $\mathbf{B}$ | $\mathbf{C}$ | $\mathbf{D}^{*}$ | S.A $^{*}$ | $\mathbf{E}$ |
| 1 | 20 | 100 | 2 | 1 | 0.5 |
| 1 | 20 | 100 | 3 | 1 | 0.5 |
| 3 | 20 | 100 | 1 | 3 | 0.2 |

## Improving the logical query plan

- Optimization with selection
- Optimization with projection
- Optimization with duplicate eliminations
- Optimization by composing the selection and cartesian product


## Optimization with selection

- Selections can be pushed down the tree as far as they can go (it reduces the size of intermediate relations and may therefore be beneficial).
- If a selection condition is the AND of several conditions, then we can split the condition and push each piece down the tree separately.


## Optimization with projection

- Projections can be pushed down the tree (it reduces the size of intermediate relations and may therefore be beneficial).
- New projections can be added.


## Optimization with duplicate eliminations

- Duplicate eliminations can be pushed down the tree as far as they can go (it reduces the size of intermediate relations and may therefore be beneficial).
- Redundant duplicate eliminations can be eliminated. Relation that is known not to have duplicates:
- A stored relation with a declared primary key
- The result of a $\gamma$ operation, since grouping creates a relation with no duplicates


## Optimization by composing the selection and cartesian product

The selection having equality as a condition can be combined with a product below to turn into an equijoin, which is generally much more efficient to evaluate than are the two operations separately.

> Database schema:
> R(A,C); S(C,D)


## Optimization by composing the selection and cartesian product

The selection having inequality as a condition can be combined with a product below to turn into an theta-join, which is generally much more efficient to evaluate than are the two operations separately.

Database schema:
R(A,B); S(C,D)


## Example of logical optimization

// IDs of suppliers which have rating greater than 50 and supply red parts.

```
SELECT
    ID_S
FROM
    S, SP, P
WHERE
    Rating > 50 AND
    S.ID_S = SP.ID_S AND
    SP.ID_P = P.ID_P AND
    Color = 'Red'
```



## Size of intermediate relations



## Pushing selections down the tree



## Pushing selections down the tree



## Pushing selections down the tree



## Replacement of selection and cartesian product by equijoin



## Size of intermediate relations

| S: | 200 tuples |
| :--- | :--- |
| SP: | 3000 tuples |
| P: | 400 tuples |
| SP x P: | $3000^{*} 400=1200000$ |
| S x SP $\times$ P: | $1200000^{*} 200=240000000$ |

- Red parts: $10 \%$
- Suppliers with rating > 50: $1 \%$
- Supplies of red parts: $20 \%$
- Suppliers supplying red parts: 50\%



## Pushing projections down the tree



