

Query Optimisation



Query Optimisation

 In practice, query optimisers incorporate elements of the following three optimisation approaches:

Semantic query optimisation

Use application specific semantic knowledge to transform a query into the one with a lower cost (they return the same answer).

Rule-based query optimisation

Use heuristic rules to transform a relational algebra expression into an equivalent one with a possibly lower cost.

Cost-based query optimisation

Use a cost model to estimate the costs of plans, and then select the most cost-effective plan.



- Can we use semantic information stored in a database (such as integrity constraints) to optimise queries?
 - semantics: "meaning".
- Recall that, integrity constraints in the relational model include:
 - key constraints
 - entity integrity constraints
 - referential integrity constraints
 - domain constraints
 - ...
 - user-defined integrity constraints
- Key idea: Integrity constraints may not only be utilized to enforce consistency of a database, but may also optimise user queries.



Example 1:

Constraint: The relation Employee has the primary key {ssn}.

Query: SELECT DISTINCT ssn FROM Employee;

 We can avoid extra costs for duplicate elimination if the existing constraint tells us that tuples in the result will be unique.



Example 2:

Constraint: No employee can earn more than 200000.

Query: SELECT name

FROM Employee

WHERE salary > 300000;

 We do not need to execute a query if the existing constraint tells us that the result will be empty.



Example 3:

Constraints: The relation WORKS_ON has the foreign keys:

[ssn]⊆EMPLOYEE[ssn] and [pno]⊆PROJECT[pnumber]

Query: SELECT DISTINCT ssn

FROM Works_on INNER JOIN Project on Works_on.pno=Project.pnumber;

 We can reduce the number of joins by executing the following query since both queries always return the same result.

> SELECT DISTINCT ssn FROM Works_on;



Rule-based Query Optimisation

- A rule-based optimisation transforms the RA expression by using a set of heuristic rules that typically improve the execution performance.
- Key ideas: apply the most restrictive operation before other operations, which can reduce the size of intermediate results:
 - Push-down selection:

Apply as early as possible to reduce the number of tuples;

- Push-down projection:
 - Apply as early as possible to reduce the number of attributes.
- Re-ordering joins:
 - Apply restrictive joins first to reduce the size of the result.
- But we must ensure that the resulting query tree gives the same result as the original query tree, i.e., the equivalence of RA expressions.

Heuristic Rules

```
Staff(sid, fname, lname, salary, position, branchNo) Branch(branchNo, name, street, suburb, city)
```

There are many heuristic rules for transforming RA expressions, utilized by the query optimiser, such as:

```
(1) \sigma_{\varphi}(\sigma_{\psi}(R)) \equiv \sigma_{\varphi \wedge \psi}(R);
\sigma_{branchNo='1'}(\sigma_{salary>60000}(Staff)) = \sigma_{branchNo='1'\wedge salary>60000}(Staff)
(2) \pi_X(\pi_Y(R)) \equiv \pi_X(R) if X \subseteq Y;
\pi_{salary}(\pi_{branchNo,salary}(Staff)) = \pi_{salary}(Staff)
(3) \sigma_{\varphi}(R_1 \times R_2) \equiv R_1 \bowtie_{\varphi} R_2
\sigma_{Staff.branchNo=Branch.branchNo}(Staff \times Branch) = (Staff) \bowtie_{Staff.branchNo=Branch.branchNo}(Branch)
```

Heuristic Rules

```
Staff(sid, fname, lname, salary, position, branchNo) Branch(branchNo, name, street, suburb, city)
```

```
(4) \ \sigma_{\varphi_{1}}(R_{1} \bowtie_{\varphi_{2}} R_{2}) \equiv R_{2} \bowtie_{\varphi_{1} \wedge \varphi_{2}} R_{1}
\sigma_{salary>60000}(Staff \bowtie_{Staff.branchNo=Branch.branchNo} (Branch)) =
(Staff) \bowtie_{Staff.branchNo=Branch.branchNo \wedge salary>60000} (Branch)
(5) \ \sigma_{\varphi}(R_{1} \bowtie R_{2}) \equiv \sigma_{\varphi}(R_{1}) \bowtie R_{2}, \text{ if } \varphi \text{ contains only attributes in } R_{1}
\sigma_{salary>60000}(Staff \bowtie_{Branch}) = \sigma_{salary>60000}(Staff) \bowtie_{Branch}
(6) \ \sigma_{\varphi_{1} \wedge \varphi_{2}}(R_{1} \bowtie R_{2}) \equiv \sigma_{\varphi_{1}}(R_{1}) \bowtie \sigma_{\varphi_{2}}(R_{2}) \text{ if } \varphi_{1} \text{ contains only attributes in } R_{1} \text{ and } \varphi_{2} \text{ contains only attributes in } R_{2}.
\sigma_{salary>60000 \wedge city='Canberra'}(Staff \bowtie_{Branch}) =
(\sigma_{salary>60000}(Staff)) \bowtie_{Gcity='Canberra'}(Branch))
```



Heuristic Rules

```
Staff(sid, fname, lname, salary, position, branchNo) Branch(branchNo, name, street, suburb, city)
```

(7) If the join condition involves only attributes in X, we have $\pi_X(R_1 \bowtie R_2) \equiv \pi_{X_1}(R_1) \bowtie \pi_{X_2}(R_2)$, where X_i contains attributes in both R_1 and R_2 , and ones in both R_i and X, and

```
\pi_{branchNo,position,city}(Staff \bowtie Branch) =
```

```
\pi_{branchNo,position}(Staff) \bowtie (\pi_{branchNo,city}(Branch))
```

(8) If the join condition contains attributes not in X, we have $\pi_X(R_1 \bowtie R_2) \equiv \pi_X(\pi_{X_1}(R_1) \bowtie \pi_{X_2}(R_2))$, where X_i contains attributes in both in R_1 and R_2 , and ones in both R_i and X

```
\pi_{position, city}(Staff \bowtie Branch) =
```

```
\pi_{position,city}(\pi_{branchNo,position}(Staff) \bowtie (\pi_{branchNo,city}(Branch)))
```

......

Push-down Selection – Example

Given the relation schemas:

```
PERSON(id, first_name, last_name, year_born)

DIRECTOR(id, title, production_year)

MOVIE_AWARD(title, production_year, award_name, year_of_award)
```

 Query: List the first and last names of the directors who have directed a movie that has won an 'Oscar' movie award

```
\pi_{\textit{first\_name}, \textit{last\_name}}(\sigma_{\textit{award\_name}='\textit{Oscar'}}((\mathsf{PERSON} \bowtie \mathsf{DIRECTOR}) \bowtie \mathsf{MOVIE\_AWARD}))
```

Question: Can we apply the following rule to optimise the query?

```
\sigma_{\varphi}(R_1 \bowtie R_2) \equiv \sigma_{\varphi}(R_1) \bowtie R_2, if \varphi contains only attributes in R_1
```



Push-down Selection – Example

Given the relation schemas:

```
PERSON(id, first_name, last_name, year_born)

DIRECTOR(id, title, production_year)

MOVIE_AWARD(title, production_year, award_name, year_of_award)
```

 Query: List the first and last names of the directors who have directed a movie that has won an 'Oscar' movie award

```
\pi_{\textit{first\_name\_iDscar'}}((\mathsf{PERSON} \bowtie \mathsf{DIRECTOR}) \bowtie \mathsf{MOVIE\_AWARD}))
```

We would have:

```
\pi_{\textit{first\_name},\textit{last\_name}}((\mathsf{PERSON} \bowtie \mathsf{DIRECTOR}) \bowtie \sigma_{\textit{award\_name}='Oscar'}(\mathsf{MOVIE\_AWARD}))
```

Push-down Projection – Example

• Given the relation schemas:

PERSON(id, first_name, last_name, year_born)

DIRECTOR(id, title, production_year)

MOVIE_AWARD(title, production_year, award_name, year_of_award)

 Query: List the first and last names of the directors who have directed a movie that has won an 'Oscar' movie award

 $\pi_{\textit{first_name},\textit{last_name}}((\mathsf{PERSON} \bowtie \mathsf{DIRECTOR}) \bowtie \sigma_{\textit{award_name}='Oscar'}(\mathsf{MOVIE_AWARD}))$

Question: Can we apply the following rule to optimise the query?

$$\pi_X(R_1\bowtie R_2)\equiv \pi_X(\pi_{X_1}(R_1)\bowtie \pi_{X_2}(R_2)),$$

where X_i contains attributes in both in R_1 and R_2 , and ones in both R_i and X



Push-down Projection – Example

• Given the relation schemas:

```
PERSON(id, first_name, last_name, year_born)

DIRECTOR(id, title, production_year)

MOVIE_AWARD(title, production_year, award_name, year_of_award)
```

 Query: List the first and last names of the directors who have directed a movie that has won an 'Oscar' movie award

```
\pi_{\textit{first\_name},\textit{last\_name}}((\mathsf{PERSON} \bowtie \mathsf{DIRECTOR}) \bowtie \sigma_{\textit{award\_name}='\mathsf{Oscar'}}(\mathsf{MOVIE\_AWARD}))
```

we would have:

```
\pi_{\textit{first\_name}, \textit{last\_name}}(\pi_{\textit{first\_name}, \textit{last\_name}, \textit{title}, \textit{production\_year}}(PERSON \bowtie DIRECTOR) \bowtie \pi_{\textit{title}, \textit{production\_year}}(\sigma_{\textit{award\_name}='Oscar'}(MOVIE\_AWARD)))
```

A Common Query Pattern (Be Careful)

- A common query pattern is join-select-project involving three steps:
 - (1) join all the relevant relations,
 - (2) select the desired tuples, and
 - (3) **project** on the required attributes.
- This query pattern can be expressed as an RA expression

$$\pi_{A_1,\ldots,A_n}(\sigma_{\varphi}(R_1\times\cdots\times R_k)),$$

or as an equivalent SQL statement

SELECT DISTINCT
$$A_1, \ldots, A_n$$
 FROM R_1, \ldots, R_k WHERE φ ;

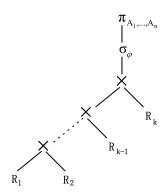
 Queries falling into this pattern can be very inefficient, which may yield huge intermediate result for the joined relations.



A Common Query Pattern (Be Careful)

push-down selection and push-down projection.

$$\pi_{A_1,\ldots,A_n}(\sigma_{\varphi}(R_1\times\cdots\times R_k)),$$





Re-ordering Joins - Example

Given the relation schemas:

```
Person(id, first_name, last_name, year_born) Suppose that it has 10000 tuples.
```

```
DIRECTOR(id, title, production_year) with [\textit{title}, \textit{production\_year}] \subseteq \mathsf{MOVIE\_AWARD}[\textit{title}, \textit{production\_year}]; \\ [\textit{id}] \subseteq \mathsf{PERSON}[\textit{id}] \text{ and } \\ \mathsf{Suppose} \text{ that it has } \textbf{100 tuples}.
```

MOVIE_AWARD(title, production_year, award_name, year_of_award) Suppose that it has **1000 tuples**.

- Example: Consider the following two RA queries. Which one is better?
 - Person ⋈ Movie_Award ⋈ Director
 - Person ⋈ Director ⋈ Movie_Award



Cost-based Query Optimisation

- A query optimiser does not depend solely on heuristic optimisation. It estimates and compares the costs of different plans.
- It estimates and compares the costs of executing a query using different execution strategies and chooses one with the lowest cost estimate.
- The query optimiser needs to limit the number of execution strategies to be considered for improving efficiency.



Summary

- In general, there are many ways of executing a query in a database.
- The user expects the result to be returned promptly, i.e., the query should be processed as fast as possible.
- But, the burden of optimising queries should not be put on the user's shoulder. The DBMSs need to do the job!
- Nonetheless, SQL is not a suitable query language in which queries can be optimised automatically.
- Instead, SQL queries are transformed into their corresponding RA queries and optimised subsequently.
- A major advantage of relational algebra is to make alternative forms of a query easy to explore.