

## Lec 15 Query Planning and Execution

### # Example

Consider  $\text{select distinct ename}$   
from Emp E, Dept D  
where E.did = D.did and D.dname = "Toy"

And catalogue with indexing info

clustered  $\rightarrow \Delta$        $\Delta$        $\Delta \leftarrow$  nonclustered  
Emp ( ssn, ename, ..., did )      10k records      1k pages  
Dept ( did, dname, ... )      500 records      50 pages

▷ Naive  $\sim 2M$  IO

$\pi_{ename}$       4 reads, 1 write  
 $\sigma_{dname = "Toy"}$       2k reads, 4 writes      (200 records if uniform)  
 $\sigma_{E.did = D.did}$       1M reads, 2k writes      (10k records)  
 $E \times D$       50 + 50k reads, 1M writes      (5M records)

▷ Ok  $\sim 54k$  IO

$\pi_{ename}$       4 reads, 1 write  
 $\sigma_{dname = "Toy"}$       2k reads, 4 writes  
 $\bowtie_{E.did = D.did}$       50 + 50k reads, 2k write [ nested loop join ]  
 $E \times D$       assume fit in buffer

△ or [ sort merge join ]  $\Rightarrow \sim 7.2k$  IO □  
□ and if vectorisation  $\Rightarrow \sim 3.2$  IO

▷ Change tree  $\sim 37$  IO

$\pi$       4 reads, 1 write  
 $\bowtie_{E.did = D.did}$       1 + 3 + 20 reads, 4 writes  
 $\sigma_{dname = "Toy"}$       E  
 $D$       [ use index ( dname ) ], 3 reads, 1 write

\* Annotated RA tree i.e. Physical Plan

↳ The tree + (access plan, estimated cardinality) per node

### # Query Optimisation

Equivalent plan space for same query is NP-hard wrt. num of joins.

Approaches:

- Heuristic, Rules
  - Push  $\sigma$  down as much as possible
  - No need to examine data
- Cost-based
  - ↳ come up with trees, estimate, pick cheapest

▷ Rules

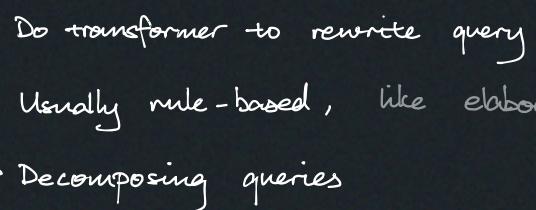
- Predicate push down
- Replace cartesian product

$$\sigma_{A.id = B.id} ( A \times B ) \rightarrow \bowtie_{A.id = B.id}$$

- Projection push down

:

### # Architecture



### # Single-rel query planning

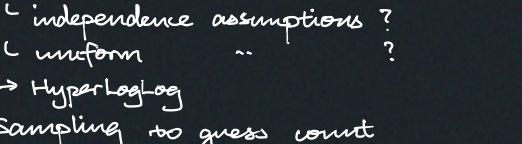
Choose good access method

### # System R query optimisation ( bottom-up approach )

1. Find best access path for each col on each table

2. Construct all possible left-deep join

left-deep      bushy



3. Find min cost among those

↳ use DP to keep best access path at each node

### # Multi-rel query optimisation ( top-down approach )

Use rules first, then do DP

Ex. we want  $[ A1 \bowtie A2 \bowtie A3 \text{ order by } A1.id ]$   $\leftarrow$  this is logical

Top-down traverse tree, apply either  $\{\text{logical} \rightarrow \text{logical}, \text{logical} \rightarrow \text{physical}\}$  rules

$[ A1 \bowtie A2 \bowtie A3 \text{ order by } A1.id ]$   $\leftarrow$  enforces rule to kill certain physical plans

$\uparrow$   $\leftarrow$  physical plan

HASH

$[ A1 \bowtie A2 ]$   $\leftarrow$  physical plan

SM

$[ A3 \bowtie A2 ]$   $\leftarrow$  physical plan

SM

$[ A1 \bowtie A3 ]$   $\leftarrow$  physical plan

SM

$[ A1 ]$   $\leftarrow$  physical plan

HASH

$[ A2 ]$   $\leftarrow$  physical plan

HASH

$[ A3 ]$   $\leftarrow$  physical plan

HASH

- + can prune without searching full space
- + can have more rules like enforcer
- harder to write

### # Rewriting

▷ Nested sub-queries rewrite

Do transformer to rewrite query

Usually rule-based, like elaborator

▷ Decomposing queries

Simply pull out nested query, and optimise top queries separately

▷ Unreachable handling

- $1 == 0 \Rightarrow \text{false}$
- $x > 5 \wedge x > 10 \Rightarrow x > 10$

### # Cost model

- Many magical numbers

- May take CPU costs

- Use internal statistics

- Build histogram

↳ allows estimating num records after range selection

→ equi-width hist. — same sized buckets

→ quantiles — about uniform buckets but diff. range

↳ independence assumptions?

↳ uniform ??

→ HyperLogLog

- Sampling to guess count