Query Optimization

Query optimization

- Given an SQL query, the query optimizer tries to figure out the order of operations that will make the query run the fastest.
- Possible because usually there is more than one way to run a query.

Why query optimization?

- SQL is a *declarative* language.
 - SQL only says *what* to retrieve from the DB, not the details of *how*.
 - Unlike most programming languages (though there are other declarative languages).
- Good query optimization can make a big difference.

Example

- Students(R#, First, Last)
- Enrolled(R#, CRN)
- SELECT First, Last FROM Students NATURAL JOIN Enrolled WHERE CRN=12345
- π_{F,L} (σ_{CRN=12345} (S⋈E))

Example

 SELECT First, Last FROM Students NATURAL JOIN Enrolled WHERE CRN=12345



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Canonical Form

- Make all JOINs explicit with WHERE clauses.
 - S NatJoin T == S Join T WHERE...
 - S Join T ON ... == S Join T WHERE...
- Perform selections and projections as early as possible.





Relational algebra

- How do we know $\pi_{F,L} (\sigma_{CRN=12345} (S \bowtie E))$ is equal to $\pi_{F,L} (S \bowtie (\sigma_{CRN=12345} (E))) ?$
- Yay 172 proofs!

What are the algorithms used?

 SELECT First, Last FROM Students NATURAL JOIN Enrolled WHERE CRN=12345



Query optimization steps

- Parse query into internal form (e.g., parse tree)
- Convert to canonical form
- Generate a set of "query plans" (a particular ordering of steps and algorithms for answering the query)
- Estimate the cost of each query plan.
- Pick the best one.

Sqlite query plan demo

- EXPLAIN QUERY PLAN
- SCAN = full table scan
- SEARCH = only a subset of the rows are visited

Back to query optimization

- Projections and selections
 - Perform them early (but carefully) to reduce
 - number of tuples
 - size of tuples (remove attributes)
 - Project out (remove) all attributes except those requested or required (e.g., needed for joins)

How does a join work?

- Three main algorithms:
 - Nested loop join
 - Sort-merge join
 - Hash join

Nested loop join

For each tuple r in R do For each tuple s in S do If r and s satisfy the join condition Then output the tuple <r,s>

Sort-Merge join

- Assume we want to join R and S on some attribute A.
- Sort both R and S by A.
- Perform two simultaneous linear scans of R and S.
 - Works well assuming no duplicate values of A.

Hash join

- Join R and S on A.
- Make a hash table of the smaller relation, mapping A to the appropriate row(s) of R (or S).
- Scan the larger relation to find the relevant rows using the hash table.
 - Only useful if smaller relation maps A to >1 rows of R.

Equivalence of expressions

• Natural joins:

 $\begin{array}{ll} - \text{ commutative} & R \vartriangleright \triangleleft S = S \rhd \triangleleft R \\ - \text{ associative} & (R \triangleright \triangleleft S) \triangleright \triangleleft T = R \triangleright \triangleleft (S \triangleright \triangleleft T) \end{array}$

• How can we figure out how many possible orderings there are to join the tables?

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 - Each join is a binary tree.
 - # of binary trees with n nodes = O(4^n) = Catalan numbers. (This only considers associativity).



Why care?

Picking good join orders

- Query optimizer generates a few potential orders
 - Doesn't evaluate all O(4ⁿ) possibilities.
 - Prefers deep trees over bushy trees. (Why?)
 - Bushy trees require lots of extra temporary tables to store intermediate results. A maximally-deep tree only requires one (or maybe two) temporary tables that we can keep overwriting.
 - How many left-deep trees are there for n relations?

- Query optimizer tries to estimate the cost for each *query plan*, relying on
 - Statistics maintained for relations and indexes (size of relation, size of index, number of distinct values in columns, etc)
 - Formulas to estimate selectivity of predicates (the probability that a randomly-selected row will be true for a predicate)
 - Formulas to estimate CPU and I/O costs of selections, projections, joins, aggregations, etc.