SQL QUERY EVALUATION

CS121: Introduction to Relational Database Systems
Fall 2014 – Lecture 12

Query Evaluation

- Last time:
 - Began looking at database implementation details
 - How data is stored and accessed by the database
 - Using indexes to dramatically speed up certain kinds of lookups
- Today: What happens when we issue a query?
 - ...and how can we make it faster?
- To optimize database queries, must understand what the database does to compute a result

Query Evaluation (2)

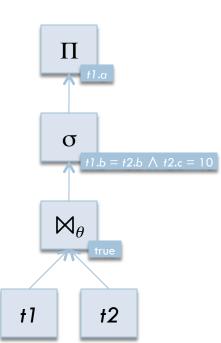
- □ Today:
 - Will look at higher-level query evaluation details
 - How relational algebra operations are implemented
 - Common-case optimizations employed in implementations
 - More details on how the database uses these details to plan and optimize your queries
- □ There are always exceptions...
 - e.g. MySQL's join processor is very different from others
 - Every DBMS has documentation about query evaluation and query optimization, for that specific database

SQL Query Processing

- Databases go through three basic steps:
 - Parse SQL into an internal representation of a plan
 - Transform this into an optimized execution plan
 - Evaluate the optimized execution plan
- Execution plans are generally based on the extended relational algebra
 - Includes generalized projection, grouping, etc.
 - Also some other features, like sorting results, nested queries, LIMIT/OFFSET, etc.

Query Evaluation Example

- A simple query:
 SELECT t1.a FROM t1, t2
 WHERE t1.b = t2.b AND t2.c = 10;
- □ Translating directly into the relational algebra: $\Pi_{t1,a}(\sigma_{t1,b} = t2,b \land t2,c = 10(t1 \times t2))$
- Database might create this structure:
 - DBs usually implement common join operations with theta-join plan nodes
 - Can be evaluated using a pushor a pull-based approach
 - $lue{}$ Evaluation loop retrieves results from top-level Π operation



Query Optimization

Are there alternate formulations of our query?

$$\Pi_{t1.a}(\sigma_{t1.b = t2.b \ \land \ t2.c = 10}(t1 \times t2))$$

$$\Pi_{t1.a}(t1 \bowtie_{t1.b = t2.b} (\sigma_{t2.c = 10}(t2)))$$

$$\Pi_{t1.a}(\sigma_{t2.c = 10}(t1 \bowtie_{t1.b = t2.b} t2))$$

- Which one is fastest?
- The query optimizer generates many equivalent plans using a set of equivalence rules
 - Cost-based optimizers assign each plan a cost, and then the lowest-cost plan is chosen for execution
 - Heuristic optimizers just follow a set of rules for optimizing a query plan

Query Evaluation Costs

- A variety of costs in query evaluation
- Primary expense is reading data from disk
 - Usually, data being processed won't fit entirely into memory
 - Try to minimize disk seeks, reads and writes!
- CPU and memory requirements are secondary
 - Some ways of computing a result require more CPU and memory resources than others
 - Becomes especially important in concurrent usage scenarios
- Can be other costs as well
 - In distributed database systems, network bandwidth must be managed by query optimizer

Query Optimization (2)

- Several questions the optimizer has to consider:
 - How is a relation's data stored on the disk?
 - ...and what access paths are available to the data?
 - What implementations of the relational algebra operations are available to use?
 - Will one implementation of a particular operation be much better or worse than another?
 - How does the database decide which query execution plan is best?
- Given the answers to these questions, what can we do to make the database go faster?

Select Operation

- \square How to implement σ_P operation?
- Easy solution from last time: scan the entire data file
 - Called a <u>file scan</u>
 - Test selection predicate against each tuple in the data file
 - Will be slow, since every disk block must be read
- This is a general solution, but not a fast one.
- What is the selection predicate P?
 - Depending on the characteristics of P, might be able to choose a more optimal evaluation strategy
 - □ If we can't, just stick with the file scan

Select Operation (2)

- Most select predicates involve a binary comparison
 - "Is an attribute equal to some value?"
 - "Is an attribute less than some value?"
- □ If data file was ordered, could use a binary search...
 - Would substantially reduce number of blocks read
 - Maintaining the logical record ordering becomes very costly if data changes frequently

Solution:

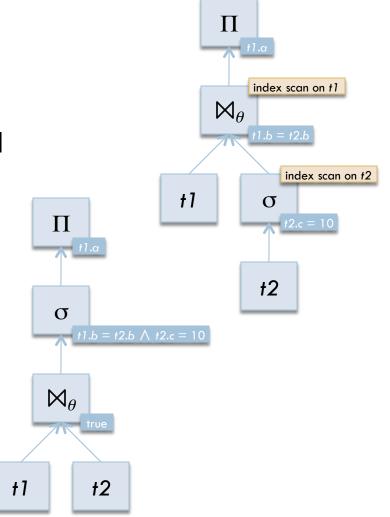
- Continue using heap file organization for table data
- For important attributes, build indexes against the data file
 - Index provides a faster way to find specific values in the data file

Select Operation

- Query planner/optimizer looks at all access paths available for a given attribute
- For select operations:
 - If select predicate is an equality test and an index is available for that attribute, can use an <u>index scan</u>
 - Can also use index scan for comparison/range tests if an ordered index is available for the attribute
- For more complicated tests, or if no index is available for attributes being used:
 - Use the simple <u>file scan</u> approach

Query Optimization Using Indexes

- Database query optimizer looks for available indexes
 - If a select/lookup operation can use an index, execution plan is annotated with this detail
 - Overall plan cost is computed including these optimizations
- Indexes can only be exploited in certain circumstances
 - Typically, only by plan nodes that directly access the table
 - e.g. original plan can't really exploit indexes at all



Project Operation

- Project operation is simple to implement
 - For each input tuple, create a new tuple with only the specified attributes
 - May also involve computed values
- Which would be faster, in general?

```
\begin{split} &\Pi_{balance}(\sigma_{balance} <_{2500}(\text{account})) \\ &\sigma_{balance} <_{2500}(\Pi_{balance}(\text{account})) \end{split}
```

- Want to project as few rows as possible, to minimize CPU and memory usage
 - Do select first: $\Pi_{balance}(\sigma_{balance < 2500}(account))$
- Good heuristic example: "Do projects as late as possible."

Sorting

- SQL allows results to be ordered
- Databases must provide sorting capabilities in execution plans
 - Data being sorted may be much larger than memory!
- For tables that fit in memory, traditional sorting techniques are used (e.g. quick-sort)
- For tables that are larger than memory, must use an external-memory sorting technique
 - Table is divided into <u>runs</u> to be sorted in memory
 - Each run is sorted, then written to a temporary file
 - All runs are merged using an N-way merge sort

Sorting (2)

- In general, sorting should be applied as late as possible
 - Ideally, rows being sorted will fit into memory
- Some other operations can also use sorted inputs to improve performance
 - Join operations
 - Grouping and aggregation
 - Usually occurs when sorted results are already available
- Could also perform sorting with an ordered index
 - Scan index, and retrieve each tuple from table file in order
 - With magnetic disks, seek-time usually makes this prohibitive
 - (solid-state disks don't have this issue!)

Join Operations

end

- Join operations are very common in SQL queries
 ...especially when using normalized schemas
 Could also potentially be a very costly operation!
 r ⋈ s defined as σ_{r,A = s,A}(r × s)
 A simple strategy for r ⋈_θ s:

 for each tuple t_r in r do begin
 if t_r, t_s satisfy condition θ then
 add t_r · t_s to result
 end
- $\Box t_r \cdot t_s$ denotes the concatenation of t_r with t_s

Nested-Loop Join

Called the nested-loop join algorithm:
 for each tuple t_r in r do begin
 for each tuple t_s in s do begin
 if t_r, t_s satisfy condition θ then
 add t_r · t_s to result
 end

- A very slow join implementation
 - Scans r once, and s once for each row in r!
 - Not so horrible if s fits entirely in memory
- But, it can handle arbitrary conditions
 - For some queries, the only option is a nested-loop join!

Indexed Nested-Loop Join

- Most join conditions involve equalities
 - Called <u>equijoins</u>
- Indexes can speed up table lookups...
- Modify nested-loop join to use indexes in inner loop:

```
for each tuple t_r in r do begin

use index on s to retrieve tuple t_s

if t_r, t_s satisfy condition \theta then

add t_r \cdot t_s to result

end
```

 Only an option for equijoins, where an index exists for the join attributes

MySQL Join Processor

- MySQL join processor is based on nested-loop join algorithm
 - Instead of joining two tables, can join N tables at once

```
for each tuple t_r in r do begin
for each tuple t_s in s do begin
for each tuple t_t in t do begin
if t_r, t_s, t_t, ... satisfy condition \theta then
add t_r \cdot t_s \cdot t_t \cdot ... to result
end
end
```

- Employs <u>many</u> optimizations
 - When possible, outer table is processed in blocks, to reduce number of iterations over inner tables
 - Indexes are exploited heavily for finding tuples in inner tables.
 - If a subquery can be resolved into a constant, it is.

MySQL Join Processor (2)

- Since MySQL join processor relies so heavily on indexes, what kinds of queries is it bad at?
 - Queries against tables without indexes... (duh)
 - Queries involving joins against derived relations (ugh!)
 - MySQL isn't smart enough to save the derived relation into a temporary table, then build an index against it
 - A common technique for optimizing complex queries in MySQL
- For more sophisticated queries, really would like more advanced join algorithms...
 - Most DBs include several other very powerful join algorithms
 - (Can't add to MySQL easily, since it doesn't use relational algebra as a query-plan representation...)

Sort-Merge Join

- If tables are already ordered by join attributes, can use a merge-sort technique
 - Must be an equijoin!
- □ Simple high-level description:
 - Two pointers to traverse tables in order:
 - p_r starts at first tuple in r
 - p_s starts at first tuple in s
 - If one pointer's tuple has join-attribute values less than the other pointer, advance that pointer
 - When pointers have the same value of the join attribute, generate joins using those rows
 - If p_r or p_s points to a run of records with the same value, must include all of these records in the join result

Sort-Merge Join (2)

- Much better performance than nested-loop join
 - Dramatically reduces disk accesses
 - Unfortunately, relations aren't usually ordered
- Can also enhance sort-merge joins when at least one relation has an index on the join attributes
 - e.g. one relation is sorted, and the unsorted relation has an index on the join attributes
 - Traverse unsorted relation's index in order
 - When rows match, use index to pull those tuples from disk
 - Disk seek cost must be managed carefully with this technique
 - e.g. can sort record pointers before reading the tuples from disk, to minimize the overall seek time

Hash Join

- Another join technique for equijoins
- \square For tables r and s:
 - Use a hash function on the join attributes to divide rows of r and s into partitions
 - \blacksquare Use same hash function on both r and s, of course
 - Partitions are saved to disk as they are generated
 - Aim for each partition to fit in memory
 - \blacksquare r partitions: $H_{r1}, H_{r2}, ..., H_{rn}$
 - \blacksquare s partitions: H_{s1} , H_{s2} , ..., H_{sn}
 - \blacksquare Rows in H_{ri} will only join with rows in H_{si}

Hash Join (2)

After partitioning: for i = 1 to n do build a hash index on H_{si} (using a second hash function!) for each row t_r in H_{ri} probe hash index for matching rows in H_{si} for each matching tuple t_s in H_{si} add $t_r \cdot t_s$ to result end end end Very fast and efficient equijoin strategy Very good for joining against derived relations!

Can perform badly when rows can't be hashed into

partitions that fit into memory

Outer Joins

- Join algorithms can be modified to generate left outer joins reasonably efficiently
 - Right outer join can be restated as left outer join
 - Will still impact overall query performance if many rows are generated
- Full outer joins can be significantly harder to implement
 - Sort-merge join can compute full outer join easily
 - Nested loop and hash join are much harder to extend
 - Full outer joins can also impact query performance heavily

Other Operations

- Set operations require duplicate elimination
 - Duplicate elimination can be performed with sorting or with hashing
- Grouping and aggregation can be implemented in several ways
 - Can sort results on the grouping attributes, then compute aggregates over the sorted values
 - All rows in a given group are adjacent to each other, so uses memory very efficiently (at least, after the sorting step...)
 - MySQL uses this approach by default
 - Can also use hashing to perform grouping and aggregation
 - Hash tuples on the grouping attributes, and compute each group's aggregate values incrementally

Optimizing Query Performance

- To improve query performance, you must know how the database actually runs your query
- Discussed the "explain" statement last time
 - Runs planner and optimizer on your query, then outputs the plan and corresponding cost estimates
- Using this information, you can:
 - Create indexes on tables, where appropriate
 - Restate the query to help the DB pick a better plan
- Harder cases may require multiple steps:
 - Generate intermediate results more well-suited for the desired query
 - Then, use intermediate results to generate final results

Query Execution Example

□ For each assignment, finds the average size of the <u>last</u> submission from students for that assignment:

```
SELECT shortname,
        AVG(last submission size) AS
             avg last submission size
FROM assignment NATURAL JOIN
      submission NATURAL JOIN
      (SELECT sub id,
               total size AS last submission size
       FROM fileset NATURAL JOIN
             (SELECT sub id, MAX(sub date) AS sub date
              FROM fileset GROUP BY sub id
               AS last sub dates
         AS last sub sizes
                               Find the date of the last fileset submitted for each
GROUP BY shortname:
                               student's submission. Name the result columns to
                               allow a natural join against the fileset table.
```

Query Execution Example (2)

 For each assignment, finds the average size of the last submission from students for that assignment:

```
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           avg last submission size
FROM assignment NATURAL JOIN
     submission NATURAL JOIN
     (SELECT sub id,
             total size AS last submission size
      FROM fileset NATURAL JOIN
           (SELECT sub id, MAX(sub date) AS sub date
            FROM fileset GROUP BY sub id
           ) AS last sub dates
        AS last sub sizes
GROUP BY shortname;
```

Join the derived result against fileset so we can retrieve the total size of the submitted files.

Query Execution Example (3)

□ For each assignment, finds the average size of the <u>last</u> submission from students for that assignment:

Outermost query finds the averages of these last submissions, and also incorporates the short-name of each assignment.

MySQL Execution and Analysis

- MySQL executes this query rather slowly*
 - About 3 sec on a server with 8GB RAM, RAID1 mirroring
 - Intuitively makes sense...
 - Joins against derived relations, non-index columns, etc.
 - All the stuff that MySQL isn't so good at handling

EXPLAIN output:

 	id	select_type	table	type	possible_keys	key	key_len			Extra
+-	1 1 1 2 2 2	PRIMARY PRIMARY PRIMARY DERIVED DERIVED DERIVED	<derived2> submission assignment <derived3> fileset submission </derived3></derived2>	eq_ref eq_ref ALL ALL		NULL PRIMARY PRIMARY NULL NULL PRIMARY	4 NULL NULL	NULL last_sub_sizes.sub_id donnie_db.submission.asn_id NULL NULL last sub dates.sub id	1506 1	
į.	3 j	DERIVED	fileset	ALL	NULL	NULL	NULL	NULL	2799	Using temporary; Using filesort

- Confirms our suspicions
- Can optimize by storing innermost results in a temp table, and creating indexes on (sub_id, sub_date)

^{*}Test was performed with MySQL 5.1; MariaDB 5.5 executes this query extremely quickly.

PostgreSQL Execution/Analysis (1)

- Postgres executes this query instantaneously. On a laptop.
 - Fundamental difference: more sophisticated join algorithms
 - Specifically hash join, which is very good at joining relations on nonindexed attributes

EXPLAIN output:

```
HashAggregate (cost=221.38..221.39 rows=1 width=8)
  -> Nested Loop (cost=144.28..221.37 rows=1 width=8)
        -> Nested Loop (cost=144.28..213.09 rows=1 width=20)
              -> Nested Loop (cost=144.28..212.81 rows=1 width=20)
                   -> Hash Join (cost=144.28..204.53 rows=1 width=12)
                         Hash Cond: ((fileset.sub id = fileset.sub id) AND ((max(fileset.sub date)) = fileset.sub date))
                         -> HashAggregate (cost=58.35..77.18 rows=1506 width=12)
                               -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=12)
                         -> Hash (cost=44.57..44.57 rows=2757 width=16)
                               -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=16)
                   -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=8)
                         Index Cond: (submission.sub id = fileset.sub id)
             -> Index Scan using assignment pkey on assignment (cost=0.00..0.27 rows=1 width=8)
                   Index Cond: (assignment.asn id = submission.asn id)
        -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=4)
             Index Cond: (submission.sub id = fileset.sub id)
```

 As expected, Postgres uses a hash join to join the derived relation against fileset table on non-index columns

PostgreSQL Execution/Analysis (2)

- □ Can disable various join algorithms in Postgres ©
 - SET enable hashjoin = off;
- EXPLAIN output:

```
HashAggregate (cost=422.68..422.69 rows=1 width=8)
  -> Nested Loop (cost=373.85..422.67 rows=1 width=8)
        -> Nested Loop (cost=373.85..414.39 rows=1 width=20)
              -> Nested Loop (cost=373.85..414.11 rows=1 width=20)
                    -> Merge Join (cost=373.85..405.83 rows=1 width=12)
                         Merge Cond: ((fileset.sub id = fileset.sub id) AND (fileset.sub date = (max(fileset.sub date))))
                         -> Sort (cost=202.12..209.01 rows=2757 width=16)
                                Sort Key: fileset.sub id, fileset.sub date
                                -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=16)
                         -> Sort (cost=171.73..175.50 rows=1506 width=12)
                                Sort Key: fileset.sub id, (max(fileset.sub date))
                               -> HashAggregate (cost=58.35..77.18 rows=1506 width=12)
                                     -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=12)
                   -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=8)
                         Index Cond: (submission.sub id = fileset.sub id)
             -> Index Scan using assignment pkey on assignment (cost=0.00..0.27 rows=1 width=8)
                   Index Cond: (assignment.asn id = submission.asn id)
        -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=4)
             Index Cond: (submission.sub id = fileset.sub id)
```

Sort + sort-merge join is still faster than nested loops!!

PostgreSQL Execution/Analysis (3)

- □ Now, disable sort-merge joins too:
 - SET enable mergejoin = off;
- Finally, Postgres performance is closer to MySQL
- EXPLAIN output:

```
HashAggregate (cost=103956.21..103956.23 rows=1 width=8)
  -> Nested Loop (cost=93.75..103956.21 rows=1 width=8)
        -> Nested Loop (cost=93.75..103947.93 rows=1 width=20)
              -> Nested Loop (cost=93.75..103947.65 rows=1 width=20)
                   -> Nested Loop (cost=93.75..103939.37 rows=1 width=12)
                          Join Filter: ((fileset.sub id = fileset.sub id) AND (fileset.sub date = (max(fileset.sub date))))
                          -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=16)
                          -> Materialize (cost=93.75..108.81 rows=1506 width=12)
                               -> HashAggregate (cost=58.35..77.18 rows=1506 width=12)
                                     -> Seq Scan on fileset (cost=0.00..44.57 rows=2757 width=12)
                   -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=8)
                         Index Cond: (submission.sub_id = fileset.sub_id)
              -> Index Scan using assignment pkey on assignment (cost=0.00..0.27 rows=1 width=8)
                   Index Cond: (assignment.asn id = submission.asn id)
       -> Index Scan using submission pkey on submission (cost=0.00..8.27 rows=1 width=4)
              Index Cond: (submission.sub_id = fileset.sub_id)
```

Query Estimates

- Query planner/optimizer must make estimates about the cost of each stage
- Database maintains statistics for each table, to facilitate planning and optimization
- Different levels of detail:
 - Some DBs only track min/max/count of values in each column. Estimates are very basic.
 - Some DBs generate and store histograms of values in important columns. Estimates are much more accurate.
- Different levels of accuracy:
 - Statistics are expensive to maintain! Databases update these statistics relatively infrequently.
 - If a table's contents change substantially, must recompute statistics

Table Statistics Analysis

- Databases also frequently provide a command to compute table statistics
- MySQL command:

```
ANALYZE TABLE assignment, submission, fileset;
```

PostgreSQL command:

```
VACUUM ANALYZE;
```

for all tables in database

```
VACUUM ANALYZE tablename;
```

- for a specific table
- These commands are expensive!
 - Perform a full table-scan
 - Also, typically lock the table(s) for exclusive access

Review

- Discussed general details of how most databases evaluate SQL queries
- Some relational algebra operations have several ways to evaluate them
 - Optimizations for very common special cases, e.g. equijoins
- Can give the database some guidance
 - Create indexes on tables where appropriate
 - Rewrite queries to be more efficient
 - Make sure statistics are up-to-date, so that planner has best chance of generating a good plan