CS 133: Databases

Fall 2019 Lec 12 – 10/15 Prof. Beth Trushkowsky

Administrivia

- Midterm this Thursday 10/17
- Assignments
 - Lab 2 ends tomorrow night, don't forget write up!
 - Lab 3 starts after fall break
 - No problem set out this week

Goals for Today

- Reason about the stages of query optimization
- Understand how to estimate the cost of a full query plan
 - Pipelining vs. materialization
 - Intermediate result sizes

Cost-based Query Sub-System



Query Optimization Overview

- Query converted to relational algebra expression
- Relational algebra converted to tree, joins as branches
- Operators can also be applied in different order!



Query Optimizer algorithm

- Goal: given a a query, the optimizer wants to
 - Decide which query plans to consider
 - Compare plans and choose the "best" one (best = shortest time to run)
- How about this algorithm?
 - Step 1: enumerate the space of all possible plans
 - Step 2: run each query plan, measure its runtime
 - Step 3: choose the plan that ran the fastest!

Query Optimizer algorithm

- Goal: given a a query, the optimizer wants to
 - Decide which query plans to consider
 - Compare plans and choose the "best" one (best = shortest time to run)
- Actual algorithm
 - Step 1: consider a set of possible plans
 - Step 2: estimate cost for each plan
 - Step 3: choose the plan with lowest cost

Estimating Cost

- Don't want to execute a plan to figure out its run-time!
 Instead estimate cost of the plan
 - Use cost as a proxy for run-time
- Cost of a plan = sum of costs for each operator in plan



Pipelined vs. Materialized

- Each query plan operator's output could be generated in either *materialized* or *pipelined* fashion
- Materialized
 - Complete output of an operator saved (typically written back to disk) as a temporary relation before its parent reads it in
- Pipelining ("on-the-fly")
 - Parts of output of operator immediately given to parent as input

Pipelining

- Parent and child operators executing concurrently
 - Iterator model
 - Parent calls next() on child/children
 - (As needed) child calls next() on its child/children
- Savings compared to materialization to disk
 - No write I/O cost for child's output
 - No read I/O cost for parent's input
- Operator algorithm(s) must support pipelining for this to work!

Exercise 3: Pipelining

- Use Page-Nested-Loop joins for the join algorithm
- Some examples:
 - (A join B) join C
 - Pipelined
 - C join (A join B)
 - Since (A join B) is the inner relation for the second join, need to materialize it





Size Estimation and Reduction Factors Statistics and cardinality estimation • Catalogs typically contain at least: SELECT attribute list • Consider a query block: - # tuples (NTuples) and # pages (NPages) per relation FROM relation list WHERE term1 AND ... AND termk and for each index: Reduction factor (RF) associated with each term # distinct key values (Nkeys) reflects the impact of the term in reducing result size low/high key values (Low/High) – Index height (Height) for each tree index. - Index size (NPages) (e.g., # leaf pages for tree) • RF is also called "selectivity" How to predict size of output? Statistics in catalogs updated periodically. - Need to know/estimate input size - Updating whenever data changes is too expensive; lots of - Need to know/estimate RFs approximation anyway, so slight inconsistency ok. - Need to know/assume how terms are related **Result Size Estimation for Selections** Exercise 6 Result cardinality (for conjunctive terms) = # input tuples * product of all RF's • RF = 16/40 * 1/10 = 1/25 - Result size: 20 pages or 1600 tuples **Assumptions:** 1. Values are uniformly distributed and *terms* are independent! 2. In System R, stats only tracked for **indexed** attributes (modern systems have removed this restriction) **Reduction Factor** Term 1 / Nkeys(I) col = valuecol > value(High(I)-value) / (High(I)-Low(I)) Note: in System R, if missing indexes, assume RF = 1/10