Query Processing

CPS 216 Advanced Database Systems

Announcements (February 17)

- * Reading assignment for this week
 - Variant indexes (due Wednesday)
- ❖ Homework #1 is being graded
 - Sample solution available outside my office
- Homework #2 due February 26
- * Midterm and course project proposal in 21/2 weeks

Overview

- ❖ Many different ways of processing the same query
 - Scan? Sort? Hash? Use an index?
 - All with different performance characteristics
- * Best choice depends on the situation
 - Implement all alternatives
 - Let the query optimizer choose at run-time

Notation

- * Relations: R, S
- \star Tuples: r, s
- * Number of tuples: |R|, |S|
- * Number of disk blocks: B(R), B(S)
- ❖ Number of memory blocks available: M
- ❖ Cost metric
 - Number of I/O's
 - Memory requirement

Table scan

- ❖ Scan table R and process the query
 - Selection over *R*
 - Projection of *R* without duplicate elimination
- **❖** I/O's: *B*(*R*)
 - Trick for selection: stop early if it is a lookup by key
- ❖ Memory requirement: 2 (double buffering)
- Not counting the cost of writing the result out
 - Same for any algorithm!
 - Maybe not needed—results may be pipelined directly into another operator

Nested-loop join

- $R\bowtie_b S$
- For each block of R, and for each r in the block: For each block of S, and for each s in the block: Output rs if p evaluates to true over r and s
 - R is called the outer table; S is called the inner table
- \bullet I/O's: $B(R) + |R| \cdot B(S)$
- * Memory requirement: 4 (double buffering)
- ❖ Improvement: block-based nested-loop join
 - For each block of R, and for each block of S:
 - For each r in the R block, and for each s in the S block: ...
 - I/O's: $B(R) + B(R) \cdot B(S)$
 - Memory requirement: same as before

More improvements of nested-loop join

- ❖ Stop early
 - If the key of the inner table is being matched
 - May reduce half of the I/O's (less for block-based)
- ❖ Make use of available memory
 - Stuff memory with as much of *R* as possible, stream *S* by, and join every S tuple with all R tuples in memory
 - I/O's: $B(R) + \left[B(R) / (M-2) \right] \cdot B(S)$
 - Or, roughly: $B(R) \cdot B(S) / M$
 - Memory requirement: M (as much as possible)

External merge sort

Problem: sort R, but R does not fit in memory

- ❖ Pass 0: read M blocks of R at a time, sort them, and write out a level-0 run
 - There are [B(R)/M] level-0 sorted runs
- ❖ Pass i: merge (M-1) level-(i-1) runs at a time, and write out a level-i run
 - (M-1) memory blocks for input, 1 to buffer output
 - # of level-i runs = $\begin{bmatrix} # \text{ of level-}(i-1) \text{ runs } / (M-1) \end{bmatrix}$
- Final pass produces 1 sorted run

Example of external merge sort

- ❖ Input: 1, 7, 4, 5, 2, 8, 9, 6, 3, 0
- * Each block holds one number, and memory has 3 blocks
- Pass 0
 - $1, 7, 4 \rightarrow 1, 4, 7$
 - 5, 2, 8 \rightarrow 2, 5, 8
 - $9, 6, 3 \rightarrow 3, 6, 9$
 - $\rightarrow 0$
- Pass 1
 - $1, 4, 7 + 2, 5, 8 \rightarrow 1, 2, 4, 5, 7, 8$
 - \bullet 3, 6, 9 + 0 \rightarrow 0, 3, 6, 9
- Pass 2 (final)
 - $1, 2, 4, 5, 7, 8 + 0, 3, 6, 9 \rightarrow 0, 1, 2, 3, 4, 5, 6, 7, 8, 9$

Performance of external merge sort

- * Number of passes: $\lceil \log_{M-1} \lceil B(R) / M \rceil \rceil + 1$
- ❖ I/O's
 - Multiply by $2 \cdot B(R)$: each pass reads the entire relation once and writes it once
 - Subtract B(R) for the final pass
 - Roughly, this is $O(B(R) \cdot \log_M B(R))$
- ❖ Memory requirement: *M* (as much as possible)

Some tricks for sorting

- Double buffering
 - Allocate an additional block for each run
 - Trade-off: smaller fan-in (more passes)
- ❖ Blocked I/O
 - Instead of reading/writing one disk block at time, read/write a bunch ("cluster")
 - Trade-off: more sequential I/O's

 smaller fan-in (more)
- * Dealing with input whose size is not an exact power of fan-in

Internal sort algorithm

- Quicksort
 - Fast
- Replacement selection
 - One block for input, one for output, rest for a heap
 - Fill the heap with input records
 - Find the smallest record in the heap that is no less than the largest record in the current run
 - If that exists, move it to the output buffer, and move a new record from input buffer into the heap
 - If that does not exist, flush output and start a new run
 - Slower than quicksort, but produces longer runs (twice the size of memory if records are in random order)

Sort-merge join

- $R\bowtie_{R.A = S.B} S$
- \diamond Sort *R* and *S* by their join attributes, and then merge r, s = the first tuples in sorted R and SRepeat until one of *R* and *S* is exhausted:

If r.A > s.B then s = next tuple in Selse if r.A < s.B then r = next tuple in Relse output all matching tuples, and r, s = next in R and S

- In most cases (e.g., join of key and foreign key)
- Worst case is $B(R) \cdot B(S)$: everything joins

❖ I/O's: sorting + 2B(R) + 2B(S)

Example

$$R: \qquad S: \qquad R \bowtie_{RA = SB} S:$$

$$\Rightarrow r_1.A = 1 \qquad \Rightarrow s_1.B = 1 \qquad r_1s_1$$

$$\Rightarrow r_2.A = 3 \qquad \Rightarrow s_2.B = 2 \qquad r_2s_3$$

$$r_3.A = 3 \qquad \Rightarrow s_3.B = 3 \qquad r_2s_4$$

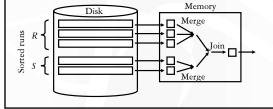
$$\Rightarrow r_4.A = 5 \qquad \Rightarrow s_4.B = 3 \qquad r_3s_3$$

$$\Rightarrow r_5.A = 7 \qquad \Rightarrow s_5.B = 8 \qquad r_3s_4$$

$$\Rightarrow r_6.A = 7 \qquad \Rightarrow r_7.s_5$$

Optimization of SMJ

- * Idea: combine join with the merge phase of merge sort
- Sort: produce sorted runs of size M for R and S
- \bullet Merge and join: merge the runs of R, merge the runs of S, and merge-join the result streams as they are generated!



Performance of two-pass SMJ

- A I/O's: $3 \cdot (B(R) + B(S))$
- * Memory requirement
 - To be able to merge in one pass, we should have enough memory to accommodate one block from each run: M >B(R) / M + B(S) / M
 - $M > \operatorname{sqrt}(B(R) + B(S))$

Other sort-based algorithms

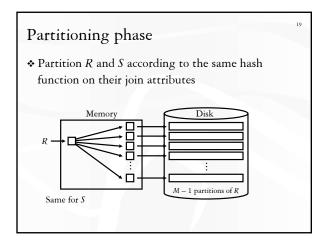
- Union (set), difference, intersection
 - More or less like SMJ
- Duplication elimination
 - External merge sort
 - · Eliminate duplicates in sort and merge
- GROUP BY and aggregation
 - External merge sort
 - Produce partial aggregate values in each run
 - Combine partial aggregate values during merge
 - · Partial aggregate values don't always work though - Examples: SUM(DISTINCT ...), MEDIAN(...)

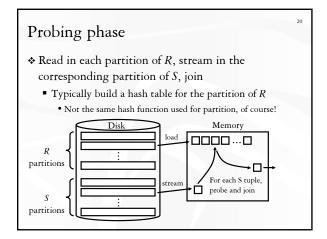
Hash join

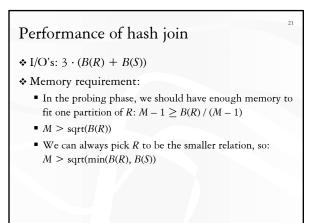
- $R\bowtie_{RA=SR} S$
- ❖ Main idea
 - Partition *R* and *S* by hashing their join attributes, and then consider corresponding partitions of R and S
 - If r.A and s.B get hashed to different partitions, they don't join

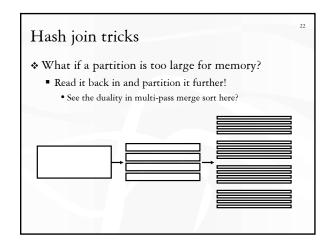


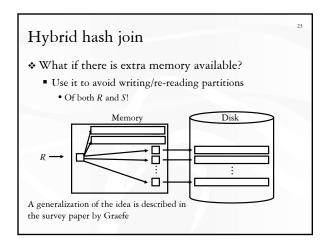
Nested-loop join considers all slots Hash join considers only those along the diagonal

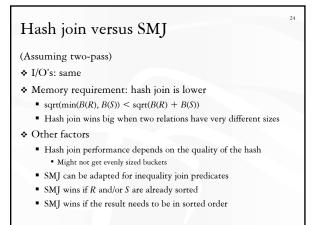












What about nested-loop join?

- * May be best if many tuples join
 - Example: non-equality joins that are not very selective
- * Necessary for black-box predicates
 - Example: ... WHERE user_defined_pred(R.A, S.B)

Other hash-based algorithms

- Union (set), difference, intersection
 - More or less like hash join
- ❖ Duplicate elimination
 - Check for duplicates within each partition/bucket
- ❖ GROUP BY and aggregation
 - Apply the hash functions to GROUP BY attributes
 - Tuples in the same group must end up in the same partition/bucket
 - Keep a running aggregate value for each group

Duality of sort and hash

* Divide-and-conquer paradigm

- Sorting: physical division, logical combination
- Hashing: logical division, physical combination
- Handling very large inputs
 - Sorting: multi-level merge
 - Hashing: recursive partitioning
- ❖ I/O patterns
 - Sorting: sequential write, random read (merge)
 - Hashing: random write, sequential read (partition)