

Evaluation of Relational Operations

Chapter 12, Part A

Relational Operations

- We will consider how to implement:
 - Selection (σ) Selects a subset of rows from relation.
 - <u>Projection</u> (π) Deletes unwanted columns from relation.
 - <u>Join</u> (>>) Allows us to combine two relations.
 - *Set-difference* (—) Tuples in reln. 1, but not in reln. 2.
 - <u>Union</u> (Y) Tuples in reln. 1 and in reln. 2.
 - Aggregation (SUM, MIN, etc.) and GROUP BY
- * Since each op returns a relation, ops can be *composed*! After we cover the operations, we will discuss how to *optimize* queries formed by composing them.

Schema for Examples

Sailors (<u>sid</u>: integer, sname: string, rating: integer, age: real) Reserves (<u>sid</u>: integer, <u>bid</u>: integer, <u>day</u>: dates, rname: string)

- Similar to old schema; rname added for variations.
- * Reserves:
 - Each tuple is 40 bytes long, 100 tuples per page, 1000 pages.
- Sailors:
 - Each tuple is 50 bytes long, 80 tuples per page, 500 pages.

Equality Joins With One Join Column

SELECT *

FROM Reserves R1, Sailors S1

WHERE R1.sid=S1.sid

- ❖ In algebra: R

 S. Common! Must be carefully optimized. R

 S is large; so,

 S followed by a selection is inefficient.
- ❖ Assume: M tuples in R, p_R tuples per page, N tuples in S, p_S tuples per page.
 - In our examples, R is Reserves and S is Sailors.
- We will consider more complex join conditions later.
- *❖ Cost metric*: # of I/Os. We will ignore output costs.

Simple Nested Loops Join

$$\label{eq:continuous_section} \begin{split} \text{for each tuple } r \text{ in } R \text{ do} \\ \text{for each tuple } s \text{ in } S \text{ do} \\ \text{if } r_i == s_j \text{ then add } < r, \, s > \text{ to result} \end{split}$$

- ❖ For each tuple in the *outer* relation R, we scan the entire *inner* relation S.
 - Cost: $M + p_R * M * N = 1000 + 100*1000*500 I/Os$.
- * Page-oriented Nested Loops join: For each *page* of R, get each *page* of S, and write out matching pairs of tuples <r, s>, where r is in R-page and S is in S-page.
 - Cost: M + M*N = 1000 + 1000*500

Index Nested Loops Join

for each tuple r in R do for each tuple s in S where $r_i == s_j$ do add <r, s> to result

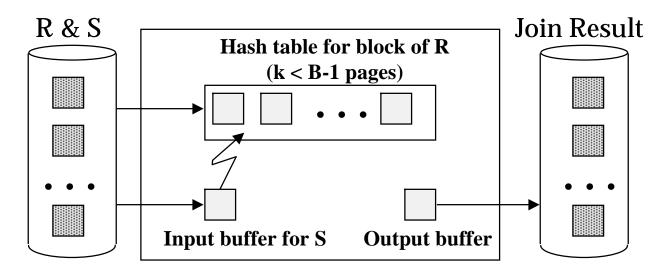
- ❖ If there is an index on the join column of one relation (say S), can make it the inner and exploit the index.
 - Cost: $M + ((M*p_R) * cost of finding matching S tuples)$
- ❖ For each R tuple, cost of probing S index is about 1.2 for hash index, 2-4 for B+ tree. Cost of then finding S tuples (assuming Alt. (2) or (3) for data entries) depends on clustering.
 - Clustered index: 1 I/O (typical), unclustered: upto 1 I/O per matching S tuple.

Examples of Index Nested Loops

- Hash-index (Alt. 2) on sid of Sailors (as inner):
 - Scan Reserves: 1000 page I/Os, 100*1000 tuples.
 - For each Reserves tuple: 1.2 I/Os to get data entry in index, plus 1 I/O to get (the exactly one) matching Sailors tuple. Total: 220,000 I/Os.
- Hash-index (Alt. 2) on sid of Reserves (as inner):
 - Scan Sailors: 500 page I/Os, 80*500 tuples.
 - For each Sailors tuple: 1.2 I/Os to find index page with data entries, plus cost of retrieving matching Reserves tuples. Assuming uniform distribution, 2.5 reservations per sailor (100,000 / 40,000). Cost of retrieving them is 1 or 2.5 I/Os depending on whether the index is clustered.

Block Nested Loops Join

- Use one page as an input buffer for scanning the inner S, one page as the output buffer, and use all remaining pages to hold ``block'' of outer R.
 - For each matching tuple r in R-block, s in S-page, add <r, s> to result. Then read next R-block, scan S, etc.



Examples of Block Nested Loops

- Cost: Scan of outer + #outer blocks * scan of inner
 - #outer blocks = [# of pages of outer / blocksize]
- ❖ With Reserves (R) as outer, and 100 pages of R:
 - Cost of scanning R is 1000 I/Os; a total of 10 blocks.
 - Per block of R, we scan Sailors (S); 10*500 I/Os.
 - If space for just 90 pages of R, we would scan S 12 times.
- With 100-page block of Sailors as outer:
 - Cost of scanning S is 500 I/Os; a total of 5 blocks.
 - Per block of S, we scan Reserves; 5*1000 I/Os.
- With <u>sequential reads</u> considered, analysis changes: may be best to divide buffers evenly between R and S.

Sort-Merge Join $(R \succeq S)$

- * Sort R and S on the join column, then scan them to do a `merge' (on join col.), and output result tuples.
 - Advance scan of R until current R-tuple >= current S tuple, then advance scan of S until current S-tuple >= current R tuple; do this until current R tuple = current S tuple.
 - At this point, all R tuples with same value in Ri (*current R group*) and all S tuples with same value in Sj (*current S group*) *match*; output <r, s> for all pairs of such tuples.
 - Then resume scanning R and S.
- * R is scanned once; each S group is scanned once per matching R tuple. (Multiple scans of an S group are likely to find needed pages in buffer.)

Example of Sort-Merge Join

sid	sname	rating	age
22	dustin	7	45.0
28	yuppy	9	35.0
31	lubber	8	55.5
44	guppy	5	35.0
58	rusty	10	35.0

sid	<u>bid</u>	day	rname
28	103	12/4/96	guppy
28	103	11/3/96	yuppy
31	101	10/10/96	dustin
31	102	10/12/96	lubber
31	101	10/11/96	lubber
58	103	11/12/96	dustin

- \diamond Cost: $M \log M + N \log N + (M+N)$
 - The cost of scanning, M+N, could be M*N (very unlikely!)
- With 35, 100 or 300 buffer pages, both Reserves and Sailors can be sorted in 2 passes; total join cost: 7500.

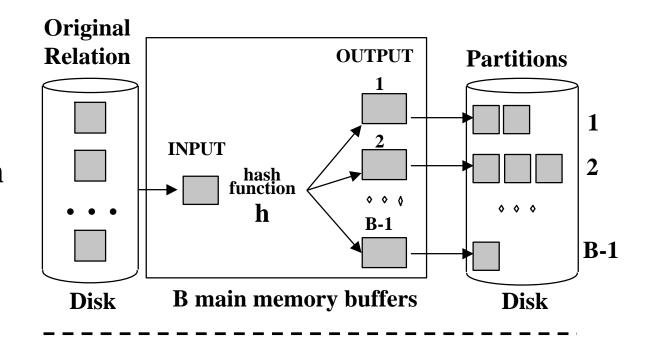
Database Management Systems, R. Ramakrishnan and Politice Cost: 2500 to 15000 I/Os)

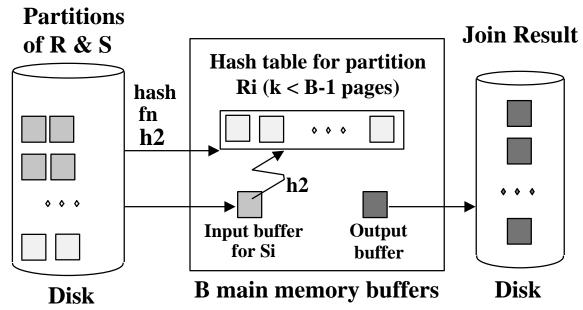
Refinement of Sort-Merge Join

- * We can combine the merging phases in the *sorting* of R and S with the merging required for the join.
 - With B > \sqrt{L} , where *L* is the size of the larger relation, using the sorting refinement that produces runs of length 2B in Pass 0, #runs of each relation is < B/2.
 - Allocate 1 page per run of each relation, and `merge' while checking the join condition.
 - Cost: read+write each relation in Pass 0 + read each relation in (only) merging pass (+ writing of result tuples).
 - In example, cost goes down from 7500 to 4500 I/Os.
- * In practice, cost of sort-merge join, like the cost of external sorting, is *linear*.

Hash-Join

- Partition both relations using hash fn h: R tuples in partition i will only match S tuples in partition i.
- Read in a partition of R, hash it using h2 (<> h!). Scan matching partition of S, search for matches.





Observations on Hash-Join

- * #partitions k < B-1 (why?), and B-2 > size of largest partition to be held in memory. Assuming uniformly sized partitions, and maximizing k, we get:
 - k= B-1, and M/(B-1) < B-2, i.e., B must be > \sqrt{M}
- * If we build an in-memory hash table to speed up the matching of tuples, a little more memory is needed.
- * If the hash function does not partition uniformly, one or more R partitions may not fit in memory. Can apply hash-join technique recursively to do the join of this R-partition with corresponding S-partition.

Cost of Hash-Join

- ❖ In partitioning phase, read+write both relns; 2(M+N). In matching phase, read both relns; M+N I/Os.
- ❖ In our running example, this is a total of 4500 I/Os.
- Sort-Merge Join vs. Hash Join:
 - Given a minimum amount of memory (*what is this, for each?*) both have a cost of 3(M+N) I/Os. Hash Join superior on this count if relation sizes differ greatly. Also, Hash Join shown to be highly parallelizable.
 - Sort-Merge less sensitive to data skew; result is sorted.

General Join Conditions

- ❖ Equalities over several attributes (e.g., R.sid=S.sid AND R.rname=S.sname):
 - For Index NL, build index on *<sid, sname>* (if S is inner); or use existing indexes on *sid* or *sname*.
 - For Sort-Merge and Hash Join, sort/partition on combination of the two join columns.
- ❖ Inequality conditions (e.g., *R.rname < S.sname*):
 - For Index NL, need (clustered!) B+ tree index.
 - Range probes on inner; # matches likely to be much higher than for equality joins.
 - Hash Join, Sort Merge Join not applicable.
 - Block NL quite likely to be the best join method here.