# Query Processing

CPS 116 Introduction to Database Systems

#### Announcements (November 9)

❖ Course project milestone #2 due today

#### Overview

- ❖ Many different ways of processing the same query
  - Scan? Sort? Hash? Use an index?
  - All have different performance characteristics and/or make different assumptions about data
- \* 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 (+1 for double buffering)
- \* Not counting the cost of writing the result out
  - Same for any algorithm!
  - Maybe not needed—results may be pipelined 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: 3 (+1 for 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
- ❖ 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:  $B(R) + \lceil B(R) / (M-2) \rceil \cdot B(S)$ 
    - Or, roughly:  $B(R) \cdot B(S) / M$
  - Memory requirement: *M* (as much as possible)
- \* Which table would you pick as the outer?

#### External merge sort

Remember (internal-memory) 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 =  $\left[ \text{# of level-}(i-1) \text{ runs } / (M-1) \right]$
- ❖ Final pass produces 1 sorted run

#### Example of external merge sort

- **❖** Input: 1, 7, 4, 5, 2, 8, 3, 6, 9
- ❖ Pass 0
  - $1, 7, 4 \rightarrow 1, 4, 7$
  - $\blacksquare$  5, 2, 8 → 2, 5, 8
  - $9, 6, 3 \rightarrow 3, 6, 9$
- ❖ Pass 1
  - $1, 4, 7 + 2, 5, 8 \rightarrow 1, 2, 4, 5, 7, 8$
  - **3**, 6, 9
- ❖ Pass 2 (final)
  - $\blacksquare 1, 2, 4, 5, 7, 8 + 3, 6, 9 \rightarrow 1, 2, 3, 4, 5, 6, 7, 8, 9$

#### 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 · 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")
  - More sequential I/O's
  - Trade-off: larger cluster → smaller fan-in (more passes)

#### Sort-merge join

- $R\bowtie_{R,A=S,B} S$
- Sort R and S by their join attributes, and then merge r, s = the first tuples in sorted R and S

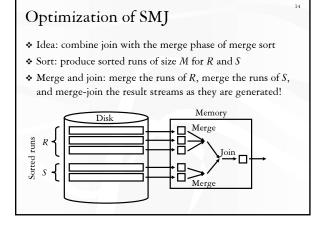
Repeat until one of R and S is exhausted:

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

- ❖ I/O's: sorting + 2 B(R) + 2 B(S)
  - In most cases (e.g., join of key and foreign key)
  - Worst case is  $B(R) \cdot B(S)$ : everything joins

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# Example R: $\Rightarrow r_1.A = 1$ $\Rightarrow s_1.B = 1$ $\Rightarrow r_2.A = 3$ $r_3.A = 3$ $\Rightarrow s_3.B = 3$ $\Rightarrow r_4.A = 5$ $\Rightarrow r_5.A = 7$ $\Rightarrow r_7.A = 8$ $R \bowtie_{R.A = S.B} S:$ $r_1.S_1$ $r_2.S_3$ $r_2.S_3$ $r_2.S_4$ $r_3.S_3$ $r_3.S_3$ $r_3.S_4$ $r_3.S_4$ $r_3.S_5$ $r_3.S_5$ $r_3.S_5$



#### Performance of two-pass SMJ

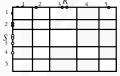
- All I/O's:  $3 \cdot (B(R) + B(S))$
- \* Memory requirement
  - To be able to merge in one pass, we should have enough memory to accommodate one block from each run: M > B(R) / M + B(S) / M
  - $M > \operatorname{sqrt}(B(R) + B(S))$

# 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

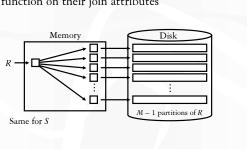
- $R\bowtie_{R.A=S} S$
- \* Main idea
  - Partition R and S by hashing their join attributes, and then consider corresponding partitions of R and S
  - If r.A and s.B get hashed to different partitions, they don't join

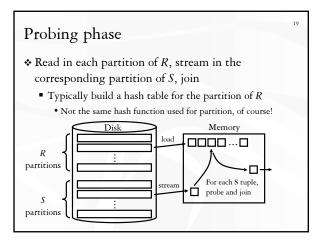


Nested-loop join considers all slots

Hash join considers only those along the diagonal

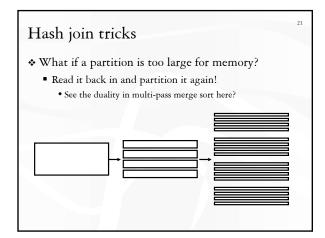
# Partitioning phasePartition R and S according to the same hash function on their join attributes





#### Performance of hash join

- $\star$  I/O's:  $3 \cdot (B(R) + B(S))$
- ❖ Memory requirement:
  - In the probing phase, we should have enough memory to fit one partition of  $R: M - 1 \ge B(R) / (M - 1)$
  - $M > \operatorname{sqrt}(B(R))$
  - We can always pick *R* to be the smaller relation, so:  $M > \operatorname{sqrt}(\min(B(R), B(S)))$



# Hash join versus SMJ

(Assuming two-pass)

- ❖ I/O's: same
- \* Memory requirement: hash join is lower
  - $\operatorname{sqrt}(\min(B(R), B(S)) < \operatorname{sqrt}(B(R) + B(S))$
  - · Hash join wins when two relations have very different sizes
- Other factors
  - Hash join performance depends on the quality of the hash
    - · Might not get evenly sized buckets
  - SMJ can be adapted for inequality join predicates
  - SMJ wins if R and/or S are already sorted
  - · SMJ wins if the result needs to be in sorted order

# 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)

#### Selection using index

**\*** Equality predicate:  $\sigma_{A=v}(R)$ 

■ Use an ISAM, B<sup>+</sup>-tree, or hash index on *R*(*A*)

\* Range predicate:  $\sigma_{A>v}(R)$ 

• Use an ordered index (e.g., ISAM or  $B^+$ -tree) on R(A)

■ Hash index is not applicable

 $\star$  Indexes other than those on R(A) may be useful

■ Example:  $B^+$ -tree index on R(A, B)

• How about  $B^+$ -tree index on R(B, A)?

#### Index versus table scan

Situations where index clearly wins:

 Index-only queries which do not require retrieving actual tuples

• Example:  $\pi_A (\sigma_{A>v}(R))$ 

\* Primary index clustered according to search key

One lookup leads to all result tuples in their entirety

#### Index versus table scan (cont'd)

BUT(!):

❖ Consider  $\sigma_{A>v}(R)$  and a secondary, non-clustered index on R(A)

Need to follow pointers to get the actual result tuples

■ Say that 20% of R satisfies A > v

• Could happen even for equality predicates

■ I/O's for index-based selection: lookup + 20% |R|

• I/O's for scan-based selection: B(R)

■ Table scan wins if a block contains more than 5 tuples

# Index nested-loop join

 $R\bowtie_{R,A=S,B} S$ 

• Idea: use the value of R.A to probe the index on S(B)

\* For each block of R, and for each r in the block: Use the index on S(B) to retrieve s with s.B = r.AOutput rs

❖ I/O's: B(R) + |R| · (index lookup)

■ Typically, the cost of an index lookup is 2-4 I/O's

• Beats other join methods if |R| is not too big

Better pick R to be the smaller relation

❖ Memory requirement: 2

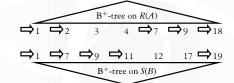
# Zig-zag join using ordered indexes

 $R\bowtie_{R,A=S,B} S$ 

**\div** Idea: use the ordering provided by the indexes on R(A) and S(B) to eliminate the sorting step of sort-merge join

\* Trick: use the larger key to probe the other index

Possibly skipping many keys that don't match



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# Summary of tricks

- Scan
  - Selection, duplicate-preserving projection, nested-loop join
- & Sor
  - External merge sort, sort-merge join, union (set), difference, intersection, duplicate elimination, GROUP BY and aggregation
- Hash
  - Hash join, union (set), difference, intersection, duplicate elimination, GROUP BY and aggregation
- ❖ Index
  - Selection, index nested-loop join, zig-zag join