Database Management and Performance Tuning Index Tuning

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Unit 6

Acknowledgements: The slides are provided by Nikolaus Augsten and adapted from "Database Tuning" by Dennis Shasha and Philippe Bonnet.

Outline

- 1 Index Tuning
 - Indexes and Joins

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Join Strategies – Running Example

- Relations: R and S
 - disk block size: 4kB
 - R: $n_r = 5000$ records, $b_r = 100$ disk blocks, 0.4MB
 - *S*: $n_s = 10000$ records, $b_s = 400$ disk blocks, 1.6*MB*
- Running Example: R ⋈ S
 - R is called the outer relation
 - S is called the inner relation

Example from Silberschatz, Korth, Sudarashan. Database System Concepts. McGraw-Hill.

Join Strategies - Naive Nested Loop

- Naive nested loop join
 - take each record of R (outer relation) and search through all records of S (inner relation) for matches
 - for each record of R, S is scanned
- Example: Naive nested loop join
 - worst case: buffer can hold only one block of each relation
 - R is scanned once, S is scanned n_r times
 - in total $n_r b_s + b_r = 2,000,100$ blocks must be read (= 8 GB)!
 - note: worst case different if S is outer relation
 - best case: both relations fit into main memory
 - $b_s + b_r = 500$ block reads

Join Strategies - Block Nested Loop

- Block nested loop join
 - compare all rows of each block of R to all records in S
 - for each block of R, S is scanned
- Example: (continued)
 - worst case: buffer can hold only one block of each relation
 - R is scanned once, S is scanned b_r times
 - in total $b_r b_s + b_r = 40,100$ blocks must be read (= 160MB)
 - best case: $b_s + b_r = 500$ block reads

Join Strategies - Indexed Nested Loop

- Indexed nested loop join
 - take each row of R and look up matches in S using index
 - runtime is $O(|R| \times \log |S|)$ (vs. $O(|R| \times |S|)$ of naive nested loop)
 - efficient if index covers join (no data access in S)
 - efficient if R has less records than S has pages: not all pages of S must be read (e.g., foreign key join from small to large table)
- Example: (continued)
 - B^+ -tree index on S has 4 layers, thus max. c=5 disk accesses per record of S
 - in total $b_r + n_r c = 25,100$ blocks must be read (= 100*MB*)

Join Strategies – Merge Join

- Merge join (two clustered indexes)
 - scan R and S in sorted order and merge
 - each block of R and S is read once
- No index on R and/or S
 - if no index: sort and store relation with $b(2\lceil log_{M-1}(b/M)\rceil + 1) + b$ block transfers (M: free memory blocks)
 - if non-clustered index present: index scan possible
- Example: (continued)
 - best case: clustered indexes on R and S (M=2 enough)
 - $b_r + b_s = 500$ blocks must be read (2MB)
 - worst case: no indexes, only M=3 memory blocks
 - sort and store R (1400 blocks) and S (7200 blocks) first: join with 9100 (36MB) block transfers in total
 - case M=25 memory blocks: 2500 block transfers (10MB)

Join Strategies - Hash Join

- Hash join (equality, no index):
 - hash both tables into buckets using the same hash function
 - join pairs of corresponding buckets in main memory
 - R is called probe input, S is called build input
- Joining buckets in main memory:
 - build hash index on one bucket from S (with new hash function)
 - probe hash index with all tuples in corresponding bucket of R
 - build bucket must fit main memory, probe bucket needs not
- Example: (continued)
 - assume that each probe bucket fits in main memory
 - R and S are scanned to compute buckets, buckets are written to disk, then buckets are read pairwise
 - in total $3(b_r + b_s) = 1500$ blocks are read/written (6MB)
 - default in SQLServer and DB2 UDB when no index present

Distinct Values and Join Selectivity

- Join selectivity:
 - number of retrieved pairs divided by cardinality of cross product $(|R \bowtie S|/|R \times S|)$
 - selectivity is low if join result is small
- Distinct values refer to join attributes of one table
- Performance decreases with number of distinct join values
 - few distinct values in both tables usually means many matching records
 - many matching records: join result is large, join slow
 - hash join: large buckets (build bucket does not fit main memory)
 - index join: matching records on multiple disk pages
 - merge join: matching records do not fit in memory at the same time

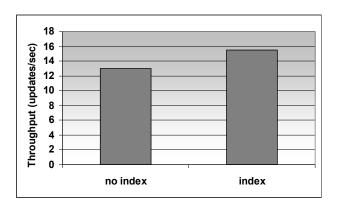
Foreign Keys

- Foreign key: attribute R.A stores key of other table, S.B
- Foreign key constraints: R.A must be subset of S.B
 - insert in R checks whether foreign key exists in S
 - deletion in S checks whether there is a record with that key in R
- Index makes checking foreign key constraints efficient:
 - index on R.A speeds up deletion from S
 - index on S.B speeds up insertion into R
 - some systems may create index on R.A and/or S.B by default
- Foreign key join:
 - each record of one table matches at most one record of the other table
 - most frequent join in practice
 - both hash and index nested loop join work well

Indexes on Small Tables

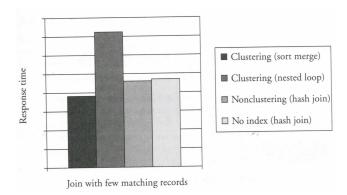
- Read query on small records:
 - tables may fit on a single track on disk
 - read query requires only one seek
 - index not useful: seeks at least one index page and one table page
- Table with large records (~page size):
 - each record occupies a whole page
 - for example, 200 records occupy 200 pages
 - index useful for point queries (read 3 pages vs. 200)
- Many inserts and deletions:
 - index must be reorganized (locking!)
 - lock conflicts near root since index is small
- Update of single records:
 - without index table must be scanned
 - scanned records are locked
 - scan (and thus lock contention) can be avoided with index

Update Queries on a Small Tables



Index avoids tables scan and thus lock contention.

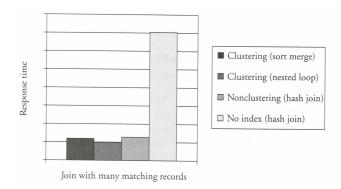
Experiment – Join with Few Matching Records



• non-clustered index is ignored, hash join used instead

SQL Server 7 on Windows 2000

Experiment – Join with Many Matching Records



- all joins slow since output size is large
- hash join (no index) slow because buckets are very large

SQL Server 7 on Windows 2000