Implementation of Relational Operations (Part 2)

R&G - Chapters 12 and 14





An Alternative to Sorting: Hashing!

• Idea:

- Many of the things we use sort for don't exploit the order of the sorted data
- e.g.: removing duplicates in DISTINCT
- e.g.: finding matches in JOIN
- Often good enough to match all tuples with equal values
- Hashing does this!
 - And may be cheaper than sorting! (Hmmm...!)
 - But how to do it for data sets bigger than memory??



General Idea

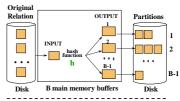
· Two phases:

- Partition: use a hash function h to split tuples into partitions on disk.
 - Key property: all matches live in the same partition.
- ReHash: for each partition on disk, build a mainmemory hash table using a hash function h2

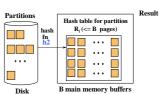


Two Phases

• Partition:



• Rehash:





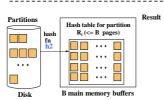
Disk B main memory buffers

Partitions Hash table for partitions R. (<= B nages)

Original Relation

 for each group of identical tuples, output one

at a time



OUTPUT

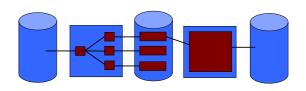
Partitions

. . .

Disk

B-1





cost = 4*[R] IO's



Memory Requirement

- · How big of a table can we hash in two passes?
 - B-1 "partitions" result from Phase 0
 - Each should be no more than B pages in size
 - Answer: B(B-1).

Said differently:

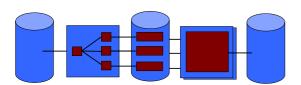
We can hash a table of size N pages in about $\sqrt{N}\,$ space

- Note: assumes hash function distributes records evenly!
- Have a bigger table? Recursive partitioning!



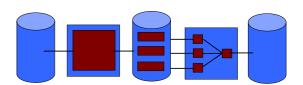
How does this compare with external sorting?





cost = 4*[R] IO's





cost = 4*[R] IO's



- How big of a table can we sort in two passes?
 - Each "sorted run" after Phase 0 is of size B
 - Can merge up to B-1 sorted runs in Phase 1
 - Answer: B(B-1).

Said differently.

We can sort a table of size N pages in about $\sqrt{N}\,$ space

• Have a bigger table? Additional merge passes!



So which is better ??

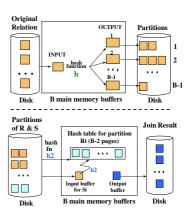
- Based on our simple analysis:
 - Same memory requirement for 2 passes
 - Same IO cost
- Digging deeper ...
- Sorting pros:
 - Great if input already sorted (or almost sorted)
 - Great if need output to be sorted anyway
 - Not sensitive to "data skew" or "bad" hash functions
- Hashing pros:
 - Highly <u>parallelizable</u> (will discuss later in semester)
 - $\bullet\,$ So is sorting, with some work
 - Can exploit extra memory to reduce # IOs (stay tuned...)



before we optimize hashing further ...

Q: Can we use hashing for JOIN?







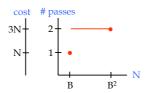
Cost of Hash Join

- Partitioning phase: read+write both relations
 ⇒ 2([R]+[S]) I/Os
- Matching phase: read both relations, write output $\Rightarrow [R]+[S]+[output]$ I/Os
- Total cost of 2-pass hash join = 3([R]+[S])+[output]
 - Q: what is cost of 2-pass sort join?
 - Q: how much memory needed for 2-pass sort join?
 - Q: how much memory needed for 2-pass hash join?



An important optimization to hashing

- Have B memory buffers
- Want to hash relation of size N

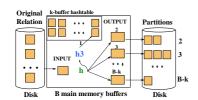


If B < N < B2, will have unused memory ...



Hybrid Hashing

• Idea: keep one of the hash buckets in memory!

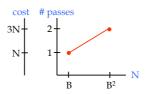


O: how do we choose the value of k?



Cost reduction due to hybrid hashing

• Now:





Summary: Hashing vs. Sorting

• Sorting pros:

- Good if input already sorted, or need output sorted
- Not sensitive to data skew or bad hash functions

• Hashing pros:

- Often cheaper due to hybrid hashing
- For join: # passes depends on size of smaller relation
- Highly parallelizable