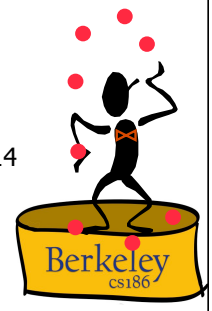
Implementation of Relational Operations (Part 2)

R&G - Chapters 12 and 14



An Alternative to Sorting: Hashing!

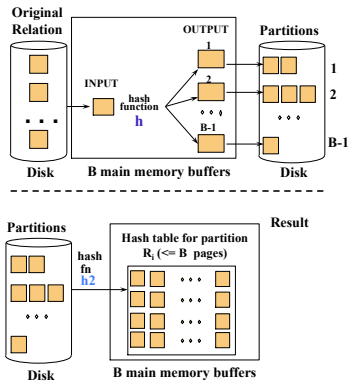
- Idea:
  - Many "sorting" tasks don't exploit the *order* of the sorted data
  - e.g.: removing duplicates in DISTINCT
  - e.g.: finding matches in JOIN
- Often good enough to match all tuples with equal values
- Hashing does this!
  - And may be cheaper than sorting! (Hmmm…!)
  - But how to do it for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

General Idea

- Two phases (divide & conquer):
  - Partition: use hash function *h* to split tuples into disk *partitions*.
    - Key property: all matches live in the same partition.
  - ReHash: for each disk partition, build a main-memory hash table using hash function *h2*

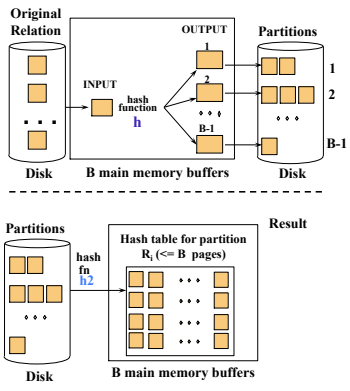
Two Phases

- Partition:
- Rehash:

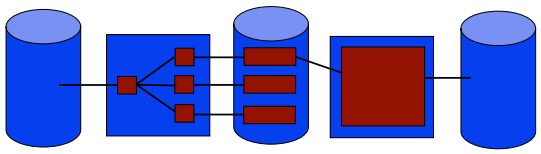


Duplicate Elimination using Hashing

- read one bucket at a time
- for each group of identical tuples, output one



Cost of External Hashing



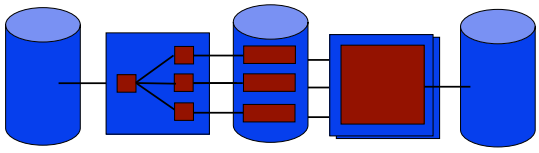
cost = 4*[R] IO's

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!

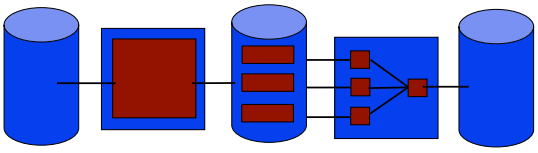
How does this compare with external sorting?

Cost of External Hashing



cost = 4*[R] IO's

Cost of External Sorting



cost = 4*[R] IO's

Memory Requirement for External Sorting

- 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 ??

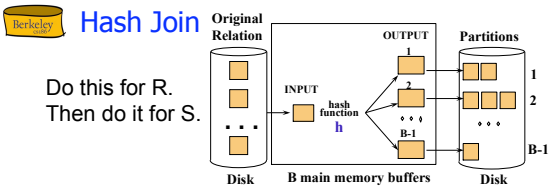
- Based on our simple analysis:
  - Same memory requirement for 2 passes
  - Same IO cost
- Digging deeper …
- Sorting pros:
  - Great if input already sorted (or *almost* sorted)
  - Great if need output to be sorted anyway
  - Not sensitive to "data skew" or "bad" hash functions
- Hashing pros:
  - For duplicate elimination/grouping, scales with # of values
    - Not # of tuples!
  - Can exploit extra memory to reduce # IOs *(stay tuned…)*

before we optimize hashing further …

Q: Can we use hashing for JOIN ?

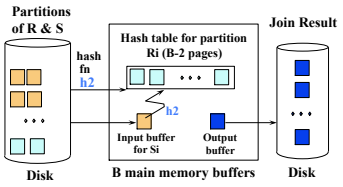
Hash Join

Do this for R.
Then do it for S.



Build an R H.T.

Stream S & *probe*.



Cost of Hash Join

- Partitioning phase: read+write both relations
  ⇒ 2([R]+[S]) I/Os
- Matching phase: read both relations, write output
  ⇒ [R]+[S] + [output] I/Os
- Total cost of 2-pass hash join = 3([R]+[S])+[output]

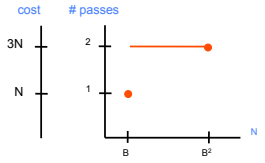
Q: what is cost of 2-pass *sort-merge join*?

Q: how much memory needed for 2-pass *sort-merge join*?

Q: how much memory needed for 2-pass *hash join*?

An important optimization to hashing

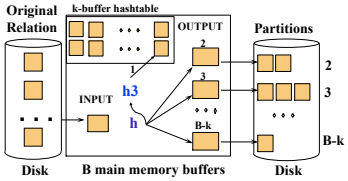
- Have B memory buffers
- Want to hash relation of size N



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

Hybrid Hashing

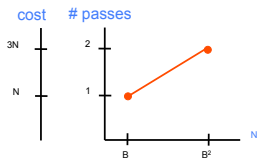
- Idea: keep one of the hash buckets in memory!



Q: 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
  - For dup-elim/grouping, depends # of values, not #tuples