

Query Processing

Sort/Hash-based (Optional)

CS348 Spring 2023

Outline

- Scan
 - Selection, duplicate-preserving projection, nested-loop join
- Index
 - Selection, index nested-loop join, zig-zag join
- Sort
 - External merge sort, sort-merge join, union (set), difference, intersection, duplicate elimination, grouping and aggregation
- Hash (Optional)

Operators that benefit from sorting

- Union (set), difference, intersection
 - More or less like SMJ
- Duplication elimination
 - External merge sort
 - Eliminate duplicates in sort and merge
- Grouping and aggregation
 - External merge sort, by group-by columns
 - Trick: produce “partial” aggregate values in each run, and combine them during merge
 - This trick doesn’t always work though
 - Examples: SUM(DISTINCT ...), MEDIAN(...)

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- Hash (Optional)
 - Hash join, union (set), difference, intersection, duplicate elimination, grouping and aggregation

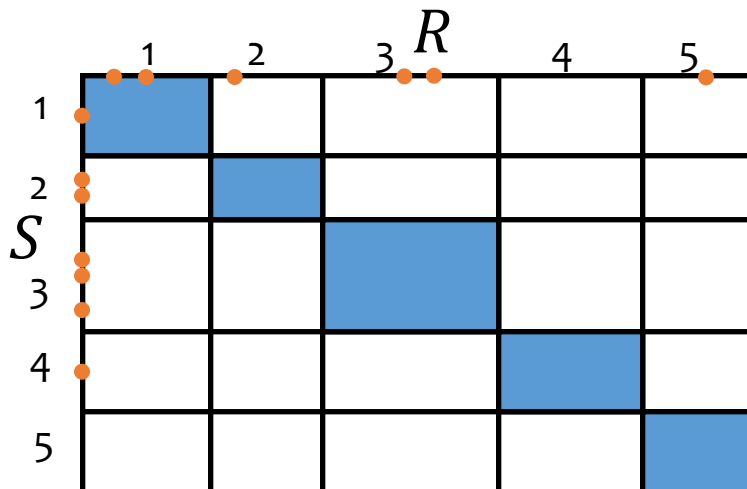
Hashing-based algorithms



Hash join

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

- 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

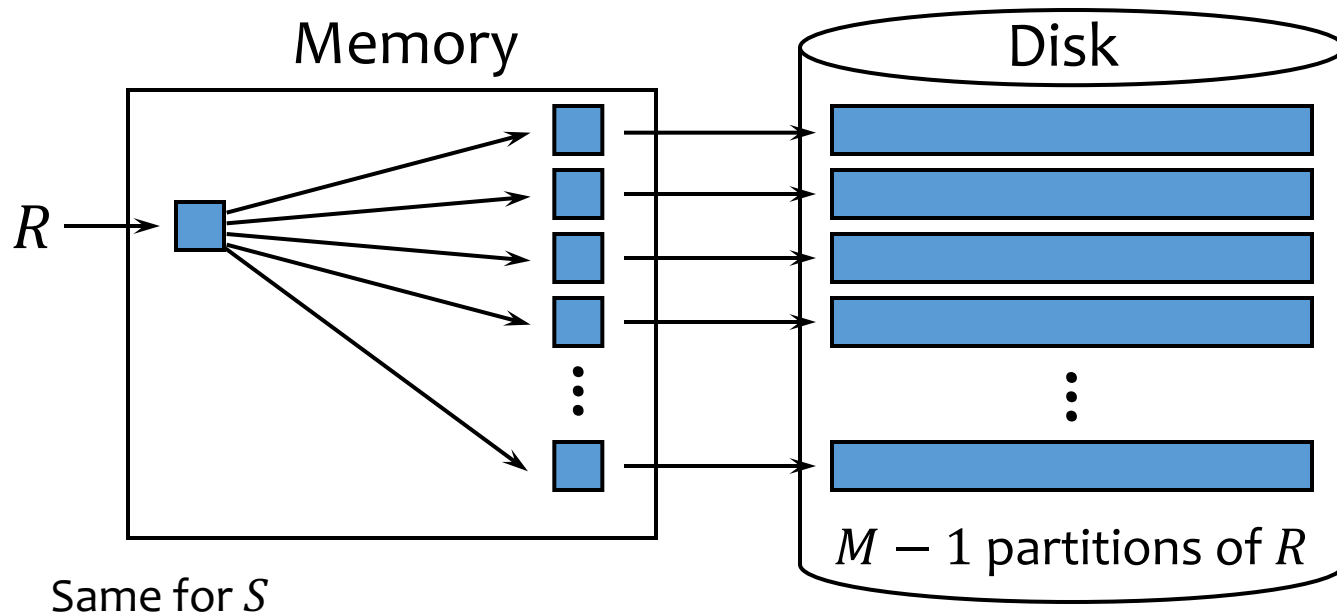


Nested-loop join
considers all slots

Hash join considers only
those along the diagonal!

Partitioning phase

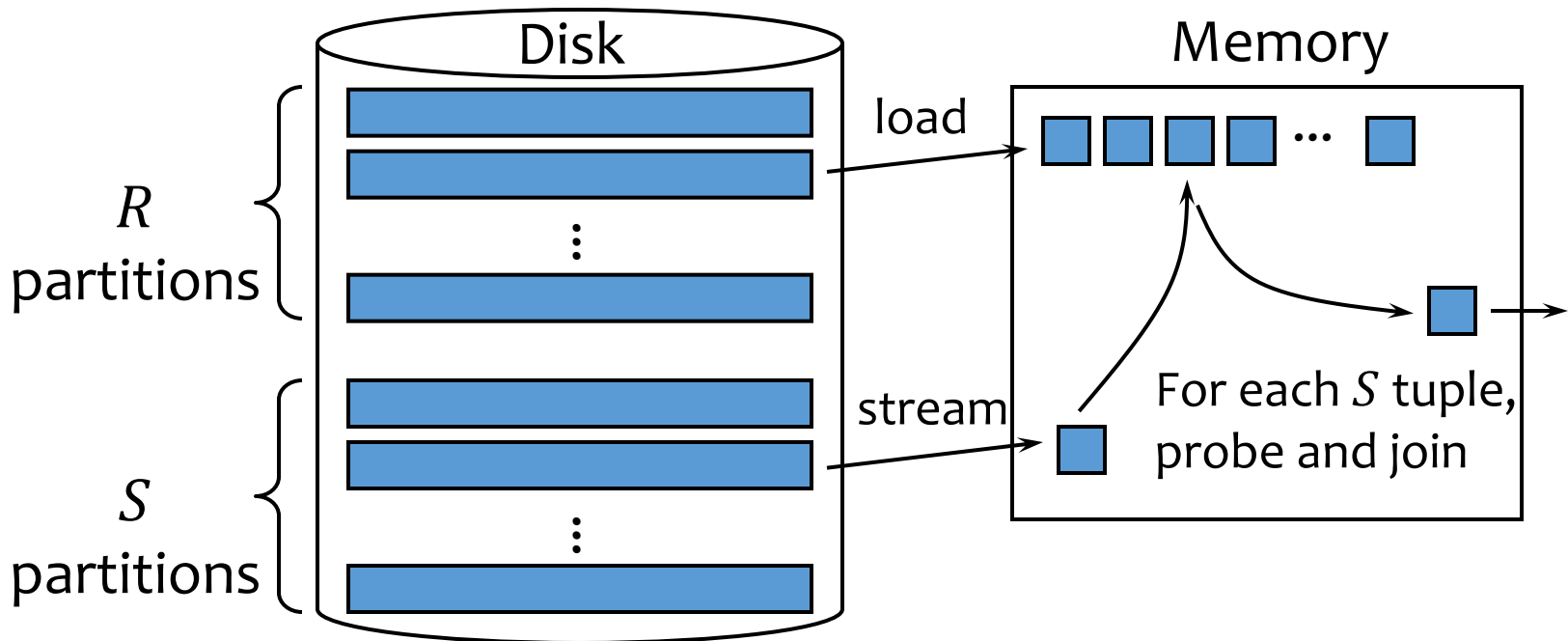
- Partition R and S according to the same hash function on their join attributes



Each partition has a size of $B(R)/(M-1)$

Probing phase

- Read in each partition of R , stream in the corresponding partition of S , join
 - Typically build a hash table for the partition of R
 - Not the same hash function used for partition, of course!



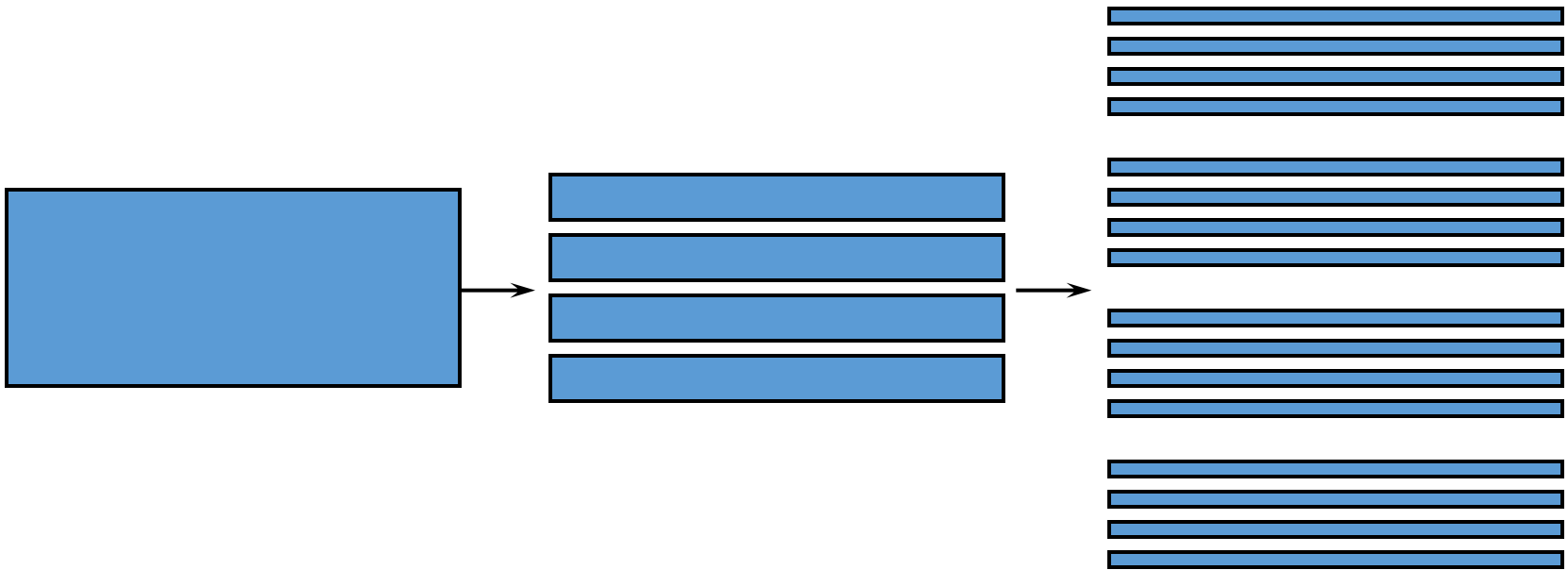
Performance of (two-pass) hash join

- If hash join completes in two phases:
 - I/O's: $3 \cdot (B(R) + B(S))$
 - 1st phase: read $B(R) + B(S)$ into memory to partition and write partitioned $B(R) + B(S)$ to disk
 - 2nd phase: read $B(R) + B(S)$ into memory to merge and join
 - Memory requirement:
 - In the probing phase, we should have enough memory to fit one partition of R : $M - 1 > \frac{B(R)}{M-1}$
 - $M > \sqrt{B(R)} + 1$
 - We can always pick R to be the smaller relation, so:

$$M > \sqrt{\min(B(R), B(S))} + 1$$

Generalizing for larger inputs

- What if a partition is too large for memory?
 - Read it back in and partition it again!
 - Re-partition $O(\log_M B(R))$ times



Hash join versus SMJ

(Assuming two-pass)

- I/O's: same
- Memory requirement: hash join is lower
 - $\sqrt{\min(B(R), B(S))} + 1 < \sqrt{B(R) + B(S)}$
 - 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
- Grouping and aggregation
 - Apply the hash functions to the group-by columns

Summary of techniques

- Scan
 - Selection, duplicate-preserving projection, nested-loop join
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- Sort (Optional)
 - External merge sort, sort-merge join, union (set), difference, intersection, duplicate elimination, grouping and aggregation
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Another view of techniques

- Selection
 - Scan without index (linear search): $O(B(R))$
 - Scan with index – selection condition must be on search-key of index
 - B+ index: $O(\log(B(R)))$
 - Hash index: $O(1)$
- Projection
 - Without duplicate elimination: $O(B(R))$
 - With duplicate elimination
 - Sorting-based: $O(B(R) \cdot \log_M B(R))$
 - Hash-based: $O(B(R) + t)$ where t is the result of the hashing phase
- Join
 - Block-based nested loop join (scan table): $O(B(R) \cdot \frac{B(S)}{M})$
 - Index nested loop join $O(B(R) + |R| \cdot (\text{index lookup}))$
 - Sort-merge join $O(B(R) \cdot \log_M B(R) + B(S) \cdot \log_M B(S))$
 - Hash join $O(B(R) \cdot \log_M B(R) + B(S) \cdot \log_M B(S))$