

CS411

Database Systems

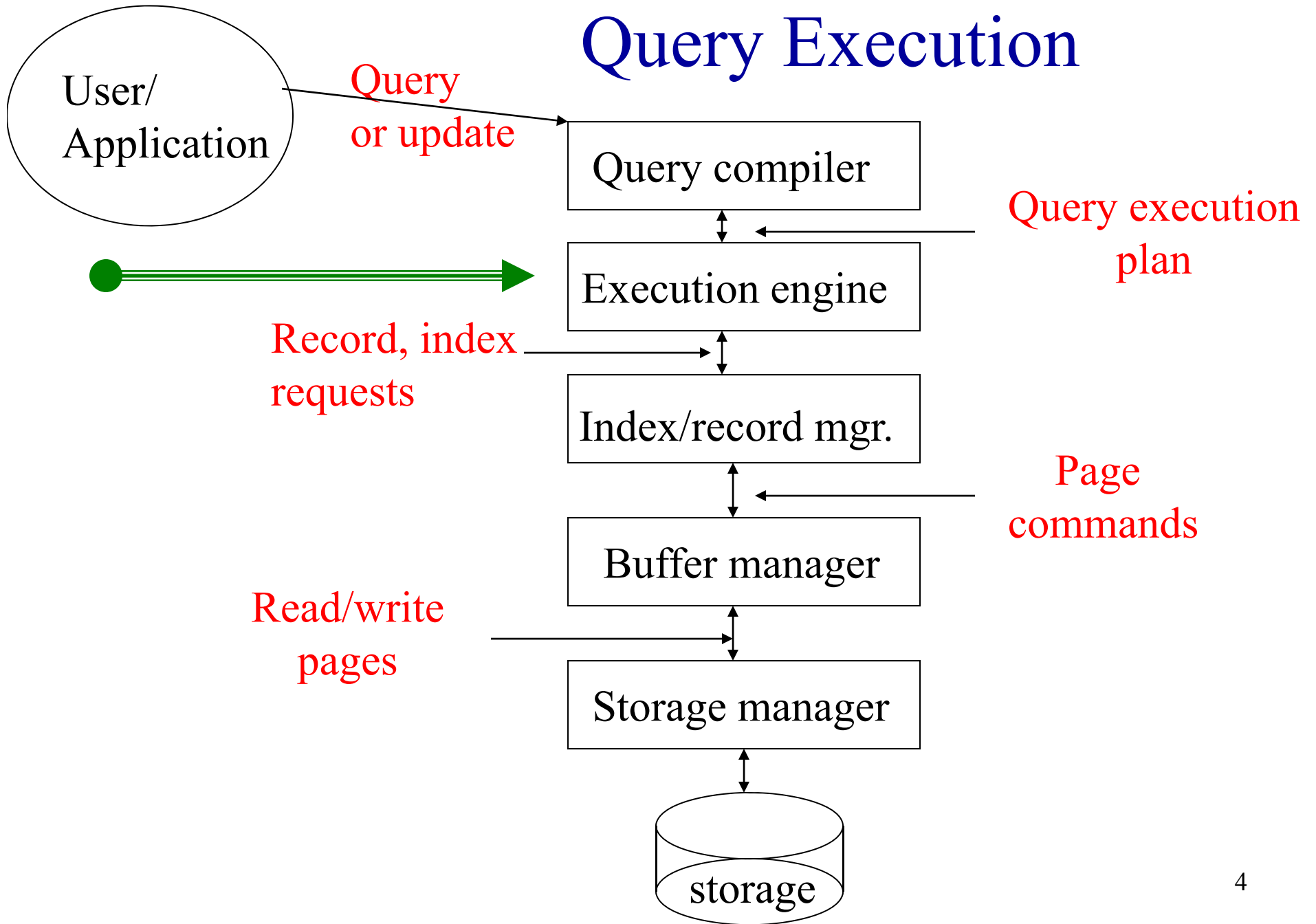
8: Query Processing

Why Do We Learn This?

Outline

- Logical/physical operators
- Cost parameters and sorting
- One-pass algorithms
- Nested-loop joins
- Two-pass algorithms

Query Execution



Logical v.s. Physical Operators

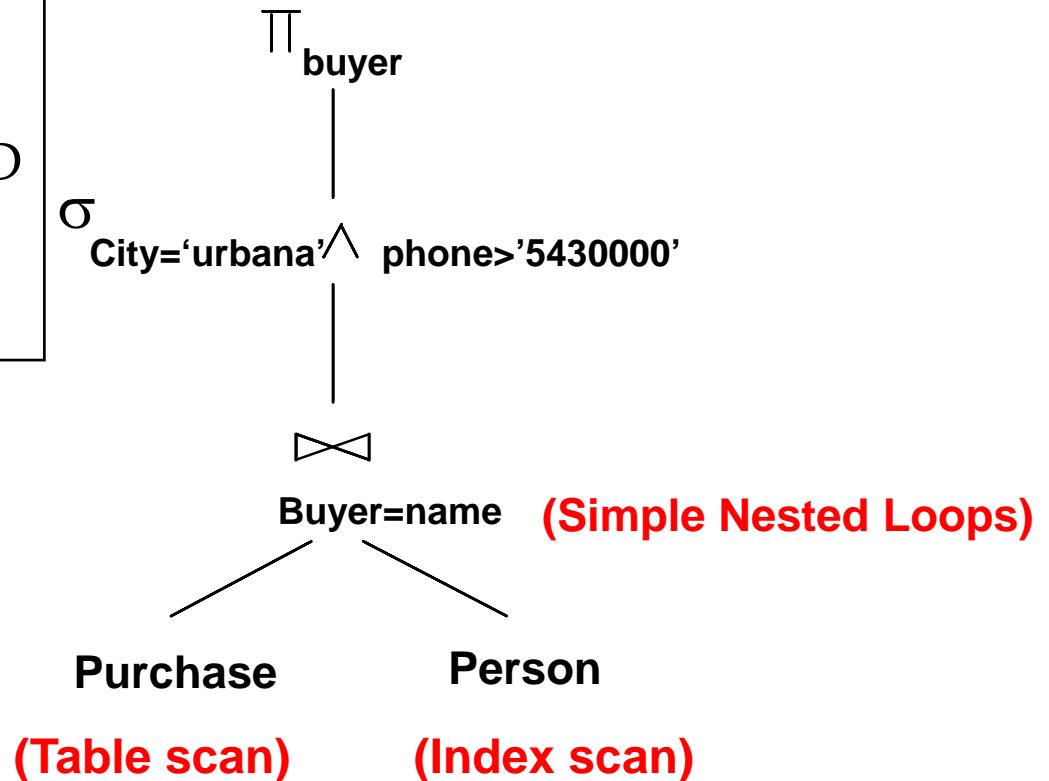
- Logical operators
 - what they do
 - e.g., union, selection, project, join, grouping
- Physical operators
 - how they do it
 - e.g., nested loop join, sort-merge join, hash join, index join

Query Execution Plans

```
SELECT S.sname
FROM   Purchase P, Person Q
WHERE  P.buyer=Q.name AND
       Q.city='urbana' AND
       Q.phone > '5430000'
```

Query Plan:

- logical tree
- implementation choice at every node
- scheduling of operations.



Some operators are from relational algebra, and others (e.g., scan, group) are not.

How do We Combine Operations?

- **The iterator model.** Each operation is implemented by 3 functions:
 - Open: sets up the data structures and performs initializations
 - GetNext: returns the the next tuple of the result.
 - Close: ends the operations. Cleans up the data structures.
- Enables pipelining!

Cost Parameters

- Cost parameters
 - M = number of blocks that fit in main memory
 - $B(R)$ = number of blocks holding R
 - $T(R)$ = number of tuples in R
 - $V(R,a)$ = number of distinct values of the attribute a
- Estimating the cost:
 - Important in optimization (next lecture)
 - Compute I/O cost only
 - We compute the cost to *read* the tables
 - We don't compute the cost to *write* the result (because pipelining)

Sorting

- Two pass multi-way merge sort
- Step 1:
 - Read M blocks at a time, sort, write
 - Result: have runs of length M on disk
- Step 2:
 - Merge $M-1$ at a time, write to disk
 - Result: have runs of length $M(M-1) \approx M^2$
- Cost: $3B(R)$, Assumption: $B(R) \leq M^2$

Scanning Tables

- The table is *clustered* (I.e. blocks consists only of records from this table):
 - Table-scan: if we know where the blocks are
 - Index scan: if we have index to find the blocks
- The table is unclustered (e.g. its records are placed on blocks with other tables)
 - May need one read for each record

Cost of the Scan Operator

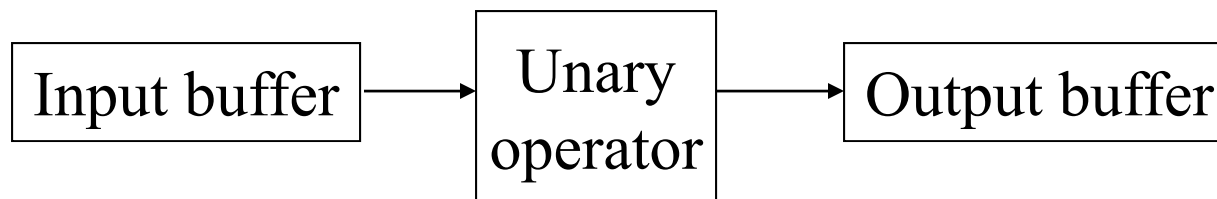
- Clustered relation:
 - Table scan: $B(R)$; to sort: $3B(R)$
 - Index scan: $B(R)$; to sort: $B(R)$ or $3B(R)$
- Unclustered relation
 - $T(R)$; to sort: $T(R) + 2B(R)$

One pass algorithms

One-pass Algorithms

Selection $\sigma(R)$, projection $\Pi(R)$

- Both are *tuple-at-a-Time* algorithms
- Cost: $B(R)$



One-pass Algorithms

Duplicate elimination $\delta(R)$

- Need to keep a dictionary in memory:
 - balanced search tree
 - hash table
 - etc
- Cost: $B(R)$
- Assumption: $B(\delta(R)) \leq M$

One-pass Algorithms

Grouping: $\gamma_{\text{city}, \text{sum}(\text{price})} (R)$

- Need to keep a dictionary in memory
- Also store the $\text{sum}(\text{price})$ for each city
- Cost: $B(R)$
- Assumption: number of cities fits in memory

One-pass Algorithms

Binary operations: $R \cap S$, $R \cup S$, $R - S$

- Assumption: $\min(B(R), B(S)) \leq M$
- Scan one table first, then the next, eliminate duplicates
- Cost: $B(R) + B(S)$

Nested Loop Joins

- Tuple-based nested loop $R \bowtie S$
- R =outer relation, S =inner relation

```
for each tuple  $r$  in  $R$  do  
    for each tuple  $s$  in  $S$  do  
        if  $r$  and  $s$  join then output  $(r,s)$ 
```

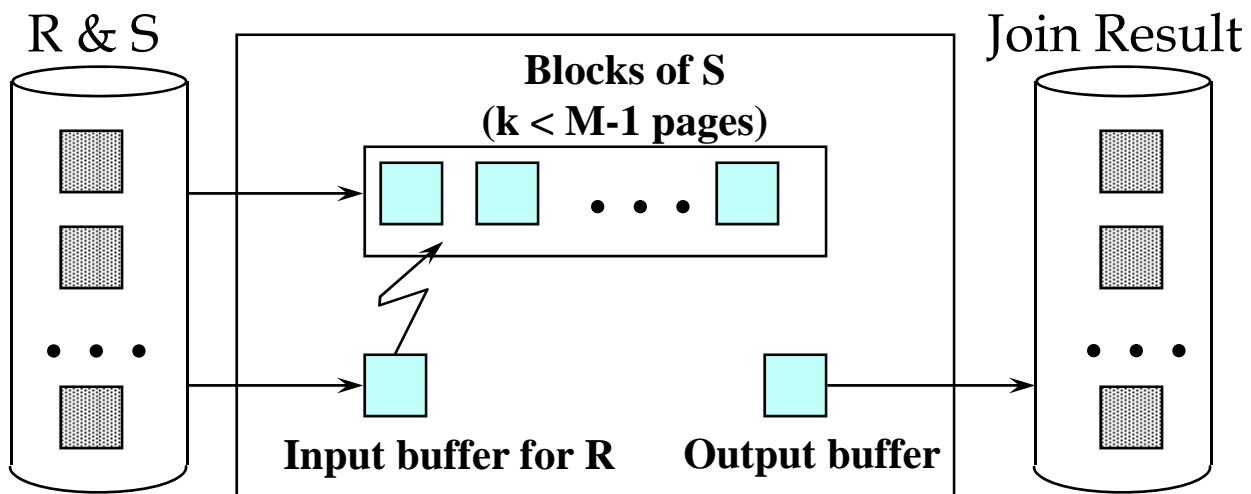
- Cost: $T(R) T(S)$, sometimes $T(R) B(S)$

Nested Loop Joins

- Block-based Nested Loop Join

```
for each (M-1) blocks bs of S do  
  for each block br of R do  
    for each tuple s in bs do  
      for each tuple r in br do  
        if r and s join then output(r,s)
```

Nested Loop Joins



Nested Loop Joins

- Block-based Nested Loop Join
- Cost:
 - Read S once: cost $B(S)$
 - Outer loop runs $B(S)/(M-1)$ times, and each time need to read R: costs $B(S)B(R)/(M-1)$
 - Total cost: $B(S) + B(S)B(R)/(M-1)$
- Notice: it is better to iterate over the smaller relation first— i.e., S smaller

Two pass algorithms

Two-Pass Algorithms Based on Sorting

Duplicate elimination $\delta(R)$

- Simple idea: like sorting, but include no duplicates
- Step 1: sort runs of size M , write
 - Cost: $2B(R)$
- Step 2: merge $M-1$ runs,
but **include each tuple only once**
 - Cost: $B(R)$
- Total cost: $3B(R)$, Assumption: $B(R) \leq M^2$

Q: What can sorting help? And, how?

- Selection?
- Projection?
- Set operations?
- Join?
- Duplicate elimination?
- Grouping?

Two-Pass Algorithms Based on Sorting

Grouping: $\gamma_{\text{city}, \text{sum}(\text{price})} (R)$

- Same as before: sort, then compute the $\text{sum}(\text{price})$ for each group
- As before: compute $\text{sum}(\text{price})$ during the merge phase.
- Total cost: $3B(R)$
- Assumption: $B(R) \leq M^2$

Two-Pass Algorithms Based on Sorting

Binary operations: $R \cap S$, $R \cup S$, $R - S$

- Idea: sort R , sort S , then do the right thing
- A closer look:
 - Step 1: split R into runs of size M , then split S into runs of size M . Cost: $2B(R) + 2B(S)$
 - Step 2: merge *all* x runs from R ; merge all y runs from S ; output a tuple on a case by cases basis ($x + y \leq M$)
- Total cost: $3B(R) + 3B(S)$
- Assumption: $B(R) + B(S) \leq M^2$

Two-Pass Algorithms Based on Sorting

Join $R \bowtie S$

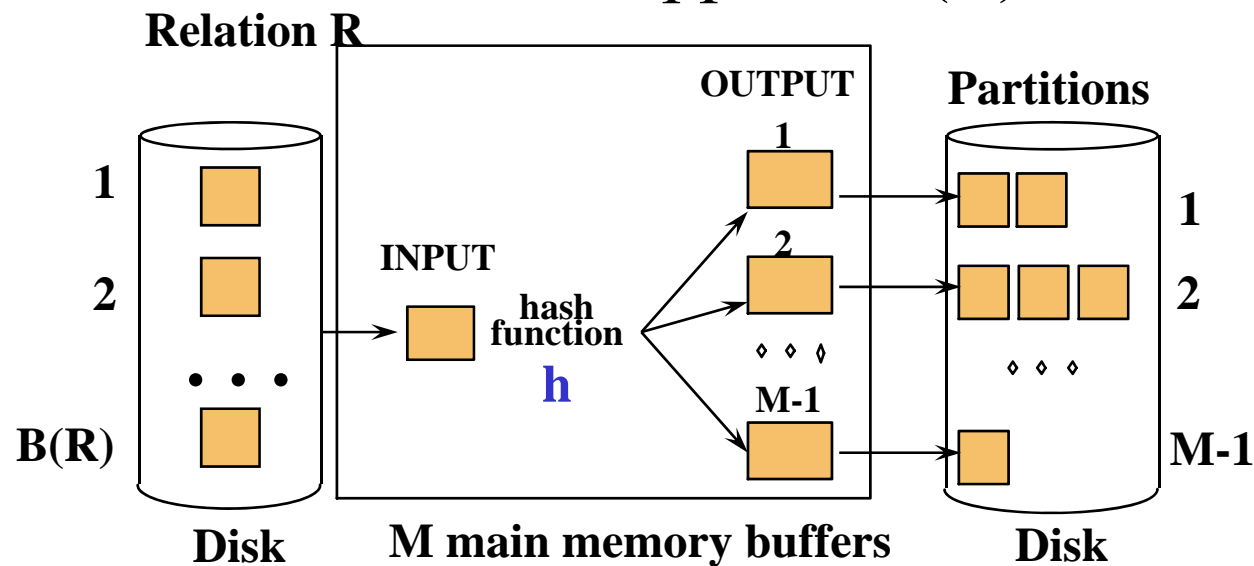
- Start by sorting both R and S on the join attribute:
 - Cost: $4B(R)+4B(S)$ (because need to write to disk)
- Read both relations in sorted order, match tuples
 - Cost: $B(R)+B(S)$
- Difficulty: many tuples in R may match many in S
 - If at least one set of tuples fits in M, we are OK
 - Otherwise need nested loop, higher cost
- Total cost: $5B(R)+5B(S)$
- Assumption: $B(R) \leq M^2, B(S) \leq M^2$

Q: Why is sorting-based “two” pass?

- Pass 1?
- Pass 2?

Two Pass Algorithms Based on Hashing

- Idea: partition a relation R into buckets, on disk
- Each bucket has size approx. $B(R)/M$



- Does each bucket fit in main memory ?
 - Yes if $B(R)/M \leq M$, i.e. $B(R) \leq M^2$

Q: What can hashing help? And, how?

- Selection?
- Projection?
- Set operations?
- Join?
- Duplicate elimination?
- Grouping?

Hash Based Algorithms for δ

- Recall: $\delta(R)$ = duplicate elimination
- Step 1. Partition R into buckets
- Step 2. Apply δ to each bucket (may read in main memory)
- Cost: $3B(R)$
- Assumption: $B(R) \leq M^2$

Hash Based Algorithms for γ

- Recall: $\gamma(R)$ = grouping and aggregation
- Step 1. Partition R into buckets
- Step 2. Apply γ to each bucket (may read in main memory)
- Cost: $3B(R)$
- Assumption: $B(R) \leq M^2$

Hash-based Join

- $R \bowtie S$
- Simple version: main memory hash-based join
 - Scan S, build buckets in main memory
 - Then scan R and join
- Requirement: $\min(B(R), B(S)) \leq M$

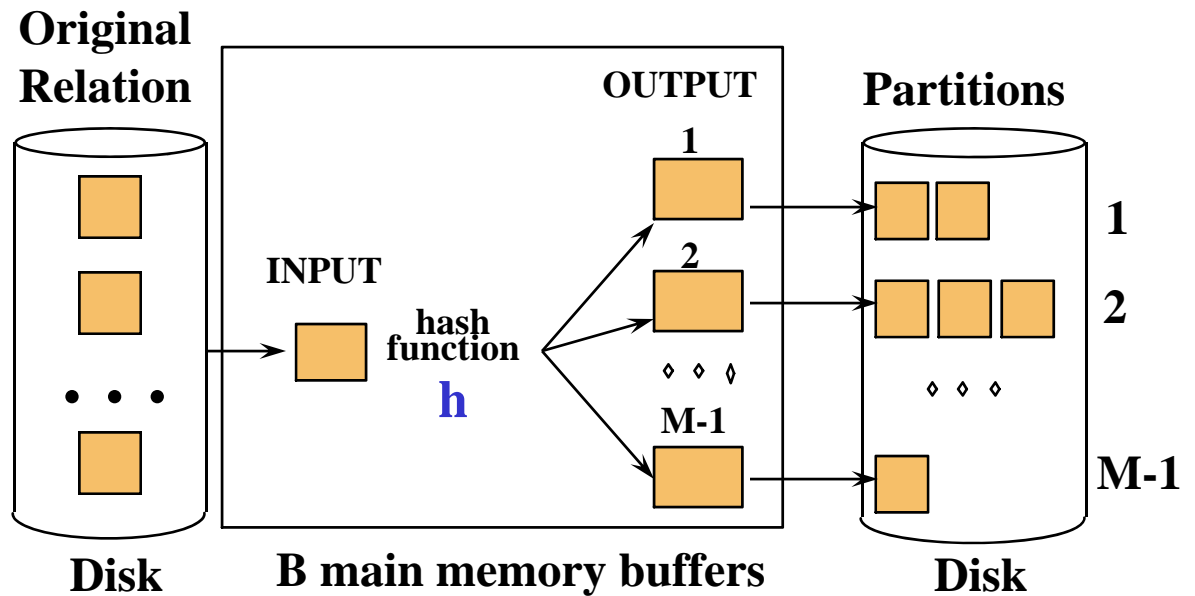
Partitioned Hash Join

$R \bowtie S$

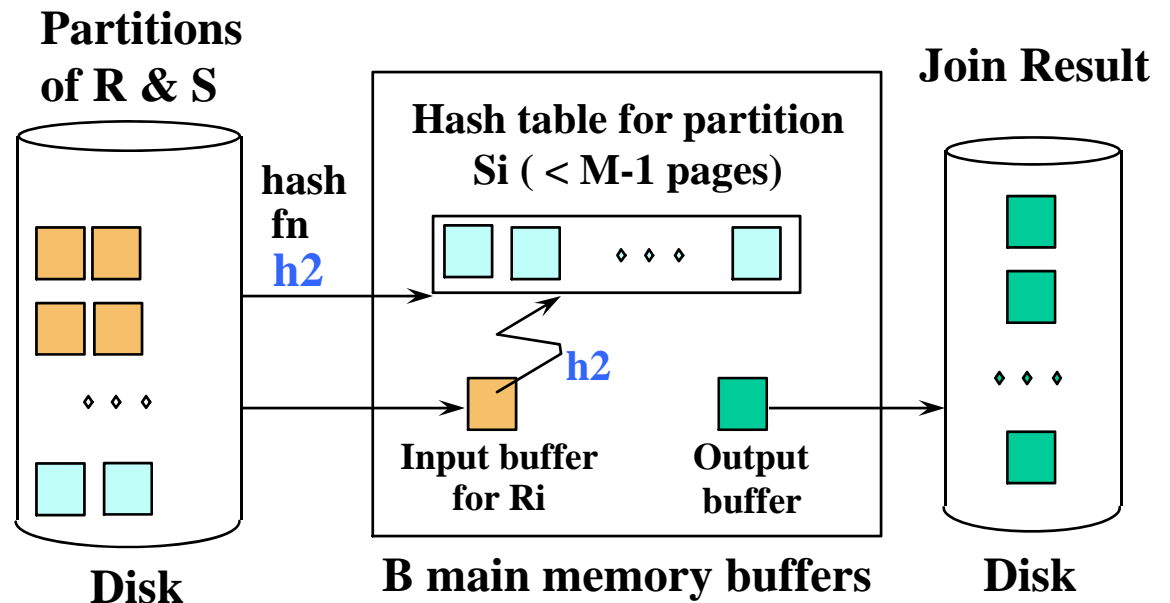
- Step 1:
 - Hash S into M buckets
 - send all buckets to disk
- Step 2
 - Hash R into M buckets
 - Send all buckets to disk
- Step 3
 - Join every pair of buckets

Partitioned Hash-Join

- Partition both relations using hash fn **h**: R tuples in partition *i* will only match S tuples in partition *i*.



- ❖ Read in a partition of R, hash it using **h2** ($\neq h$). Scan matching partition of S, search for matches.



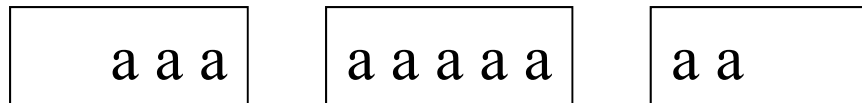
Partitioned Hash Join

- Cost: $3B(R) + 3B(S)$
- Assumption:
At least one full bucket of the smaller rel must fit in
memory: $\min(B(R), B(S)) \leq M^2$

Index-based algorithms (zero-pass!)

Indexed Based Algorithms

- In a clustered index all tuples with the same value of the key are clustered on as few blocks as possible



Index Based Selection

- Selection on equality: $\sigma_{a=v}(R)$
- Clustered index on a: cost $B(R)/V(R,a)$
- Unclustered index on a: cost $T(R)/V(R,a)$

Index Based Selection

- Example: $B(R) = 2000$, $T(R) = 100,000$, $V(R, a) = 20$, compute the cost of $\sigma_{a=v}(R)$
- Cost of table scan:
 - If R is clustered: $B(R) = 2000$ I/Os
 - If R is unclustered: $T(R) = 100,000$ I/Os
- Cost of index based selection:
 - If index is clustered: $B(R)/V(R,a) = 100$
 - If index is unclustered: $T(R)/V(R,a) = 5000$
- Notice: when $V(R,a)$ is small, then unclustered index is useless

Index Based Join

- $R \bowtie S$
- Assume S has an index on the join attribute
- Iterate over R , for each tuple fetch corresponding tuple(s) from S
- Assume R is clustered. Cost:
 - If index is clustered: $B(R) + T(R)B(S)/V(S,a)$
 - If index is unclustered: $B(R) + T(R)T(S)/V(S,a)$

Index Based Join

- Assume both R and S have a sorted index (B+ tree) on the join attribute
- Then perform a merge join (called zig-zag join)
- Cost: $B(R) + B(S)$