CSE 544 Principles of Database Management Systems

Magdalena Balazinska Winter 2009 Lecture 7 - Query execution and operator algorithms

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References

- Join processing in database systems with large main memories. Leonard Shapiro. ACM Transactions on Database Systems 11(3), 1986. Also in Red Book (3rd and 4th ed)
- The Anatomy of a Database System. J. Hellerstein and M. Stonebraker. Section 4. Red Book. 4th Ed.
- Database management systems.
 Ramakrishnan and Gehrke.
 Third Ed. Chapters 12, 13 and 14.

Outline

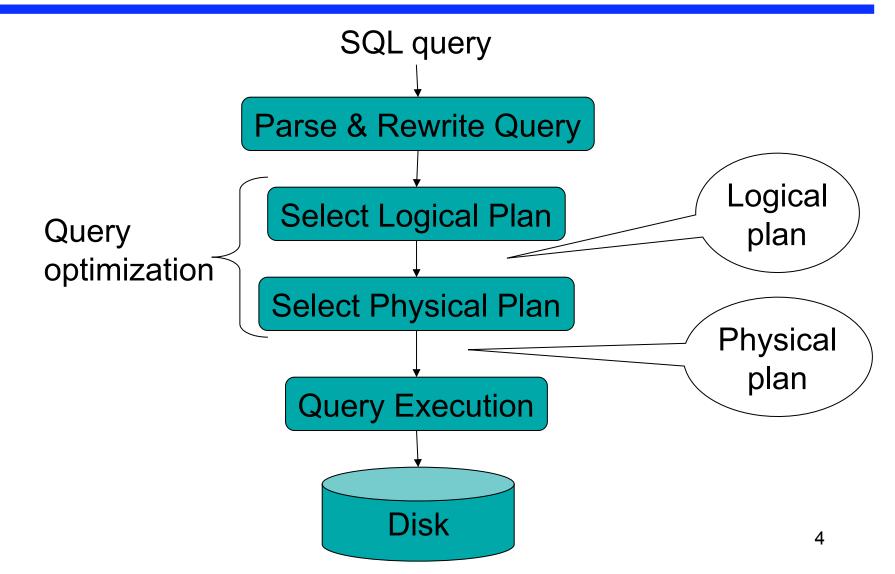
Steps involved in processing a query

- Logical query plan
- Physical query plan
- Query execution overview

Operator implementations

- One pass algorithms
- Two-pass algorithms
- Index-based algorithms

Query Evaluation Steps



Example Database Schema

Supplier(sno, sname, scity, sstate)
Part(pno, pname, psize, pcolor)
Supply(sno, pno, price)

View: Suppliers in Seattle

CREATE VIEW NearbySupp AS SELECT sno, sname FROM Supplier WHERE scity='Seattle' AND sstate='WA'

Example Query

 Find the names of all suppliers in Seattle who supply part number 2

SELECT sname FROM NearbySupp WHERE sno IN (SELECT sno FROM Supplies WHERE pno = 2)

Steps in Query Evaluation

• Step 0: admission control

- User connects to the db with username, password
- User sends query in text format

Step 1: Query parsing

- Parses query into an internal format
- Performs various checks using catalog
 - Correctness, authorization, integrity constraints

• Step 2: Query rewrite

- View rewriting, flattening, etc.

Rewritten Version of Our Query

Original query:

```
SELECT sname
FROM NearbySupp
WHERE sno IN ( SELECT sno
FROM Supplies
WHERE pno = 2 )
```

Rewritten query:

```
SELECT S.sname
FROM Supplier S, Supplies U
WHERE S.scity='Seattle' AND S.sstate='WA'
AND S.sno = U.sno
AND U.pno = 2;
```

Continue with Query Evaluation

• Step 3: Query optimization

- Find an efficient query plan for executing the query
- We will spend a whole lecture on this topic

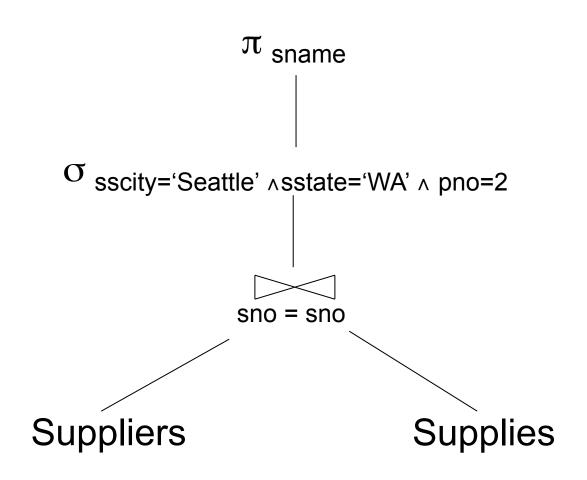
• A query plan is

- Logical query plan: an extended relational algebra tree
- Physical query plan: with additional annotations at each node
 - Access method to use for each relation
 - Implementation to use for each relational operator

Extended Algebra Operators

- Union ∪, intersection ∩, difference -
- Selection or
- Projection π
- Join 🖂
- Duplicate elimination δ
- Grouping and aggregation $\boldsymbol{\gamma}$
- Sorting τ
- Rename ρ

Logical Query Plan

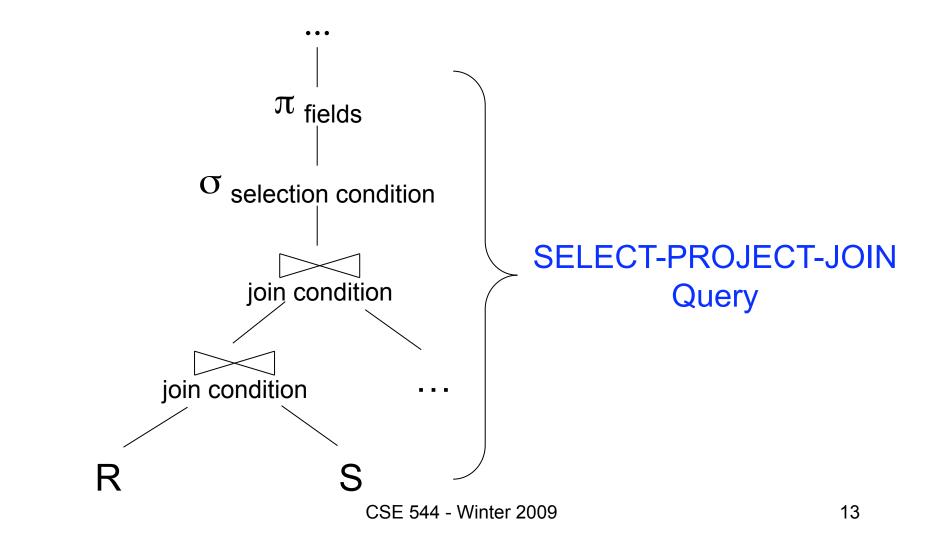


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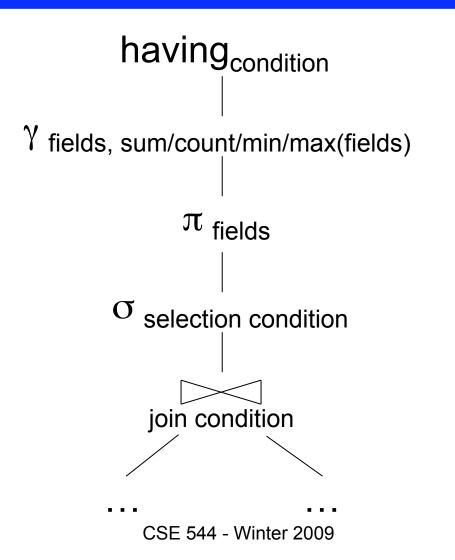
Query Block

- Most optimizers operate on individual query blocks
- A query block is an SQL query with **no nesting**
 - Exactly one
 - SELECT clause
 - FROM clause
 - At most one
 - WHERE clause
 - GROUP BY clause
 - HAVING clause

Typical Plan for Block (1/2)



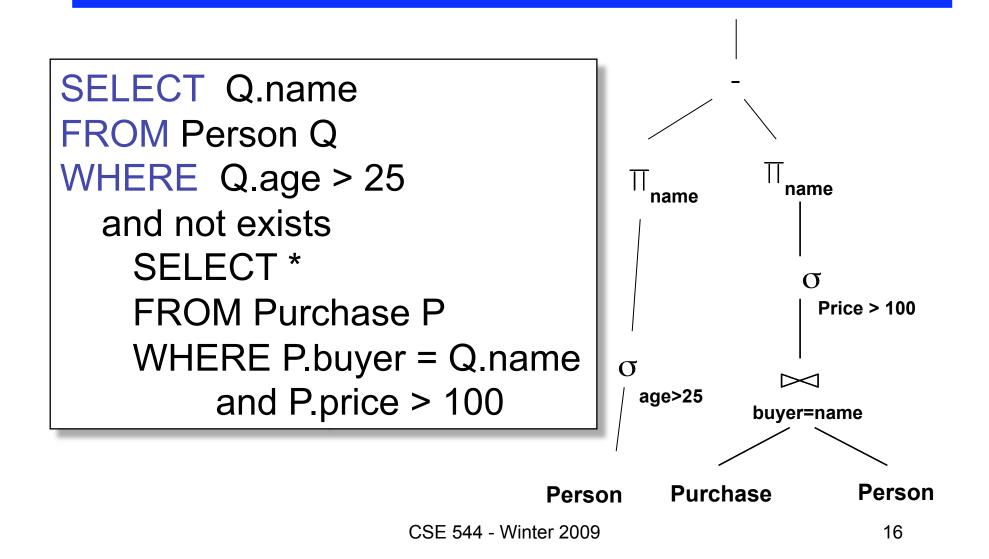
Typical Plan For Block (2/2)



How about Subqueries?

```
SELECT Q.name
FROM Person Q
WHERE Q.age > 25
and not exists
SELECT *
FROM Purchase P
WHERE P.buyer = Q.name
and P.price > 100
```

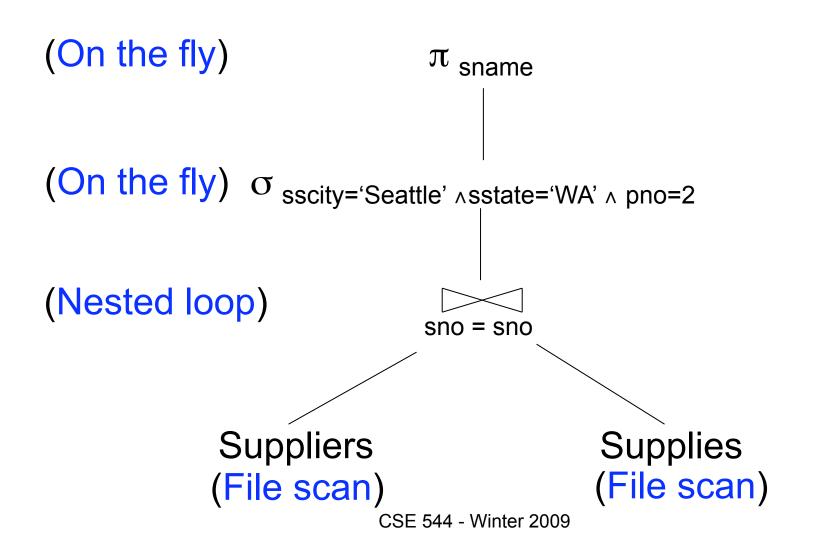
How about Subqueries?



Physical Query Plan

- Logical query plan with extra annotations
- Access path selection for each relation
 - Use a file scan or use an index
- Implementation choice for each operator
- Scheduling decisions for operators

Physical Query Plan



Final Step in Query Processing

Step 4: Query execution

- How to synchronize operators?
- How to pass data between operators?
- What techniques are possible?
 - One thread per process
 - Iterator interface
 - Pipelined execution
 - Intermediate result materialization

Iterator Interface

- Each operator implements this interface
- Interface has only three methods
- open()
 - Initializes operator state
 - Sets parameters such as selection condition
- get_next()
 - Operator invokes get_next() recursively on its inputs
 - Performs processing and produces an output tuple
- close(): clean-up state

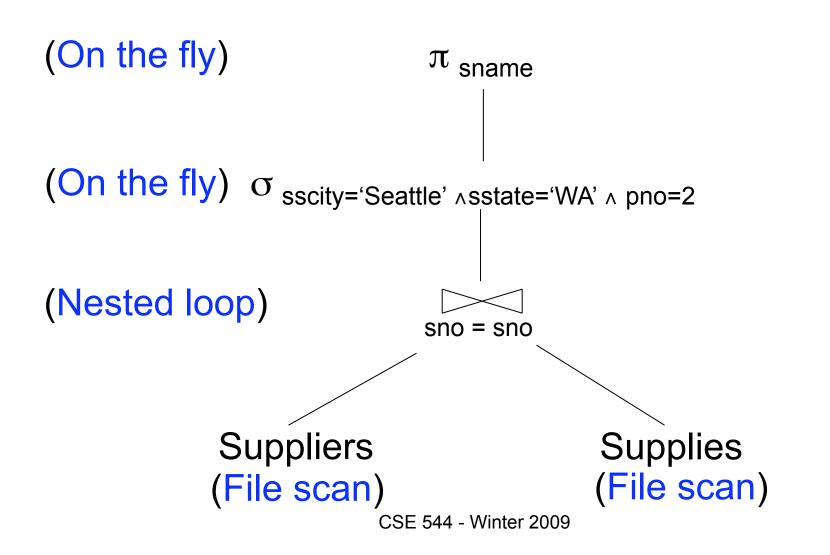
Pipelined Execution

 Applies parent operator to tuples directly as they are produced by child operators

Benefits

- No operator synchronization issues
- Saves cost of writing intermediate data to disk
- Saves cost of reading intermediate data from disk
- Good resource utilizations on single processor
- This approach is used whenever possible

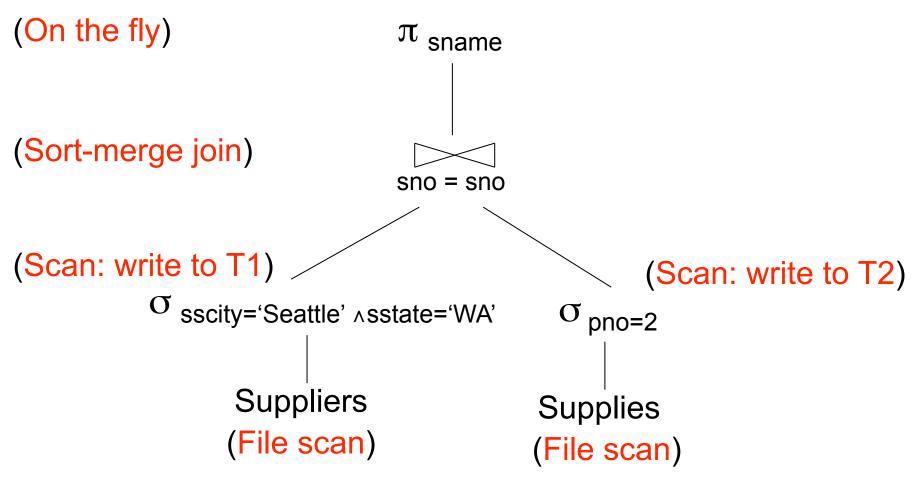
Pipelined Execution



Intermediate Tuple Materialization

- Writes the results of an operator to an intermediate table on disk
- No direct benefit but
- Necessary for some operator implementations
- When operator needs to examine the same tuples multiple times

Intermediate Tuple Materialization



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Outline

• Steps involved in processing a query

- Logical query plan
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- Query execution overview

Operator implementations

- One pass algorithms
- Two-pass algorithms
- Index-based algorithms

Why Learn About Op Algos?

- Implemented in commercial DBMSs
- Different DBMSs implement different subsets of these algorithms
- Good algorithms can greatly improve performance
- Need to know about physical operators to understand query optimization

Cost Parameters

- In database systems the data is on disk
- Cost = total number of I/Os
- Parameters:
 - B(R) = # of blocks (i.e., pages) for relation R
 - T(R) = # of tuples in relation R
 - V(R, a) = # of distinct values of attribute a

Cost

- Cost of an operation = number of disk I/Os to
 - read the operands
 - compute the result
- Cost of writing the result to disk is *not included*
 - Need to count it separately when applicable

Notions of Clustering

- **Clustered-file organization** (aka co-clustering)
 - Tuples of one relation R are placed with a tuple of another relation S with a common value

Clustered relation

- Tuples of relation are stored on blocks predominantly devoted to storing that relation
- Sometimes also called "clustered file organization"
- Clustered index (aka clustering index)
 - When ordering of data records is close to the ordering of data entries in the index

Cost Parameters

- Clustered relation R:
 - Blocks consists mostly of records from this table
 - $B(R) \approx T(R) / blockSize$
- Unclustered relation R:
 - Its records are placed on blocks with other tables
 - When R is unclustered: $B(R) \approx T(R)$
- When a is a key, V(R,a) = T(R)
- When a is not a key, V(R,a)

Cost of Scanning a Table

- Clustered relation:
 - Result may be unsorted: B(R)
 - Result needs to be sorted: 3B(R)
- Unclustered relation
 - Unsorted: T(R)
 - Sorted: T(R) + 2B(R)

One-pass Algorithms

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

- Both are *tuple-at-a-time* algorithms
- Cost: B(R), the cost of scanning the relation



Join Algorithms

- Logical operator:
- Propose three physical operators for the join, assuming the tables are in main memory:
 - Hash join
 - Nested loop join
 - Sort-merge join

Hash Join

Hash join: $R \bowtie S$

- Scan R, build buckets in main memory
- Then scan S and join
- Cost: B(R) + B(S)
- One pass algorithm when B(R) <= M

Nested Loop Joins

- Tuple-based nested loop R ⋈ S
- R is the outer relation, S is the inner relation

<u>for</u> each tuple r in R <u>do</u> <u>for</u> each tuple s in S <u>do</u> <u>if</u> r and s join <u>then</u> output (r,s)

- Cost: B(R) + T(R) B(S) when S is clustered
- Cost: B(R) + T(R) T(S) when S is unclustered

Page-at-a-time Refinement

for each page of tuples r in R do for each page of tuples s in S do for all pairs of tuples if r and s join then output (r,s)

- Cost: B(R) + B(R)B(S) if S is clustered
- Cost: B(R) + B(R)T(S) if S is unclustered

- We can be much more clever
- How would you compute the join in the following cases ?
 What is the cost ?

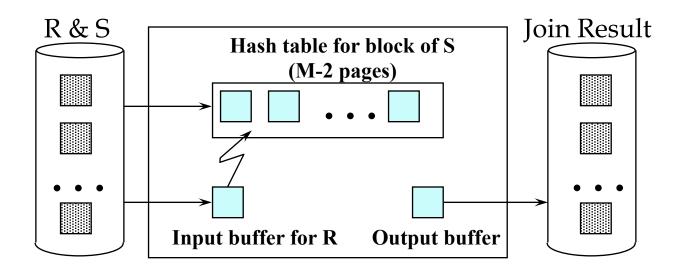
$$- B(R) = 1000, B(S) = 2, M = 4$$

- B(R) = 1000, B(S) = 3, M = 4

$$-$$
 B(R) = 1000, B(S) = 6, M = 4

- Block Nested Loop Join
- Group of (M-2) pages of S is called a "block"

```
for each (M-2) pages ps of S do
for each page pr of R do
for each tuple s in ps
for each tuple r in pr do
if "r and s join" then output(r,s)
```



- Cost of block-based nested loop join
 - Read S once: cost B(S)
 - Outer loop runs B(S)/(M-2) times, and each time need to read R: costs B(S)B(R)/(M-2)
 - Total cost: B(S) + B(S)B(R)/(M-2)
- Notice: it is better to iterate over the smaller relation first

Sort-Merge Join

Sort-merge join: $R \bowtie S$

- Scan R and sort in main memory
- Scan S and sort in main memory
- Merge R and S
- Cost: B(R) + B(S)
- One pass algorithm when B(S) + B(R) <= M
- Typically, this is NOT a one pass algorithm

One-pass Algorithms

Duplicate elimination $\delta(R)$

- Need to keep tuples in memory
- When new tuple arrives, need to compare it with previously seen tuples
- Balanced search tree or hash table
- Cost: B(R)
- Assumption: $B(\delta(R)) \le M$

One-pass Algorithms

Grouping:

Product(name, department, quantity)

 $\gamma_{department, sum(quantity)}$ (Product) \rightarrow Answer(department, sum)

How can we compute this in main memory ?

One-pass Algorithms

- Grouping: $\gamma_{department, sum(quantity)}$ (R)
- Need to store all departments in memory
- Also store the sum(quantity) for each department
- Balanced search tree or hash table
- Cost: B(R)
- Assumption: number of depts fits in memory

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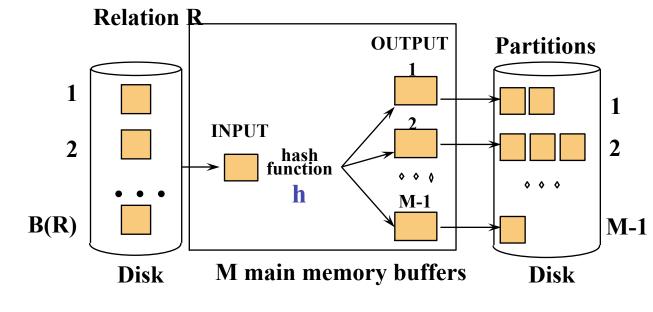
- One pass algorithms
- Two-pass algorithms
- Index-based algorithms

Two-Pass Algorithms

- What if data does not fit in memory?
- Need to process it in multiple passes
- Two key techniques
 - Hashing
 - Sorting

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 ?

Hash Based Algorithms for δ

- Recall: $\delta(R)$ = duplicate elimination
- Step 1. Partition R into buckets
- Step 2. Apply δ to each bucket
- Cost: 3B(R)
- Assumption: B(R) <= M²

Hash Based Algorithms for $\,\gamma$

- Recall: $\gamma(R)$ = grouping and aggregation
- Step 1. Partition R into buckets
- Step 2. Apply γ to each bucket
- Cost: 3B(R)
- Assumption: B(R) <= M²

Simple Hash Join

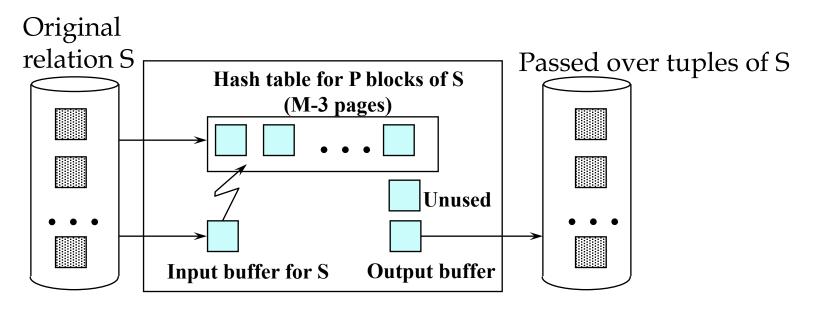
 $\mathsf{R}\bowtie\mathsf{S}$

- Step 1:
 - P = min(M-3, B(S))
 - Choose hash function h and set of hash values s.t. P blocks of S tuples will hash into that set
 - Hash S and either insert tuple into hash table or write to disk
- Step 2
 - Hash R and either probe the hash table for S or write to disk
- Step 3
 - Repeat steps 1 and 2 until all tuples are processed

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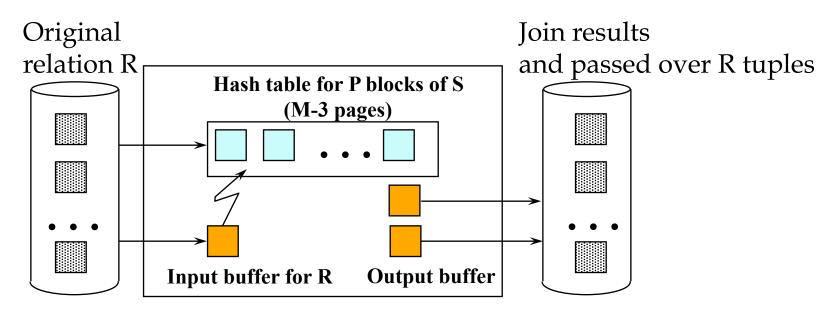
Simple Hash Join

- Build a hash-table for M-3 pages of S
- Write remaining pages of S back to disk



Simple Hash Join

- Hash R using the same hash function
- Probe hash table for S or write tuples of R back to disk



- Repeat these two steps until all tuples are processed
- Requires many passes

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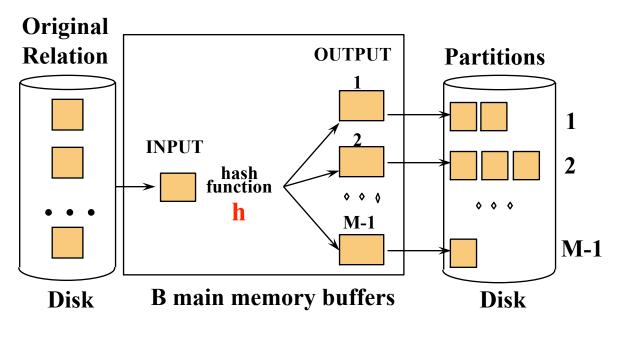
Partitioned (Grace) Hash Join

$\mathsf{R}\bowtie\mathsf{S}$

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

Partitioned Hash Join

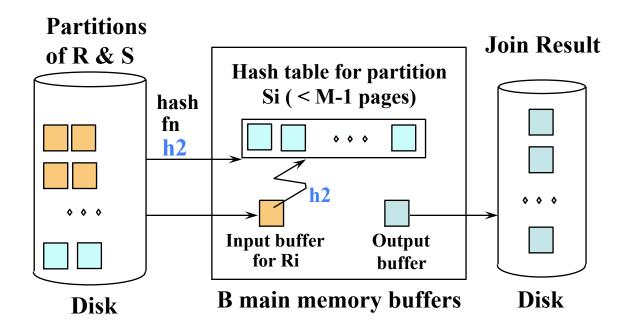
- Partition both relations using hash fn h
- R tuples in partition i will only match S tuples in partition i.



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Partitioned Hash Join

- Read in partition of R, hash it using h2 (\neq h)
 - Build phase
- Scan matching partition of S, search for matches
 - Probe phase



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Partitioned Hash Join

- Cost: 3B(R) + 3B(S)
- Assumption: min(B(R), B(S)) <= M²

Hybrid Hash Join Algorithm

- Assume we have extra memory available
- Partition S into k buckets

 t buckets S₁, ..., S_t stay in memory
 k-t buckets S_{t+1}, ..., S_k to disk
- Partition R into k buckets
 - First t buckets join immediately with S
 - Rest k-t buckets go to disk
- Finally, join k-t pairs of buckets:

 $(R_{t+1}, S_{t+1}), (R_{t+2}, S_{t+2}), ..., (R_k, S_k)$

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Hybrid Hash Join Algorithm

- How to choose k and t?
 - Choose k large but s.t.
 - Choose t/k large but s.t.
 - Moreover:

k <= M t/k * B(S) <= M t/k * B(S) + k-t <= M

• Assuming t/k * B(S) >> k-t: t/k = M/B(S)

Hybrid Hash Join Algorithm

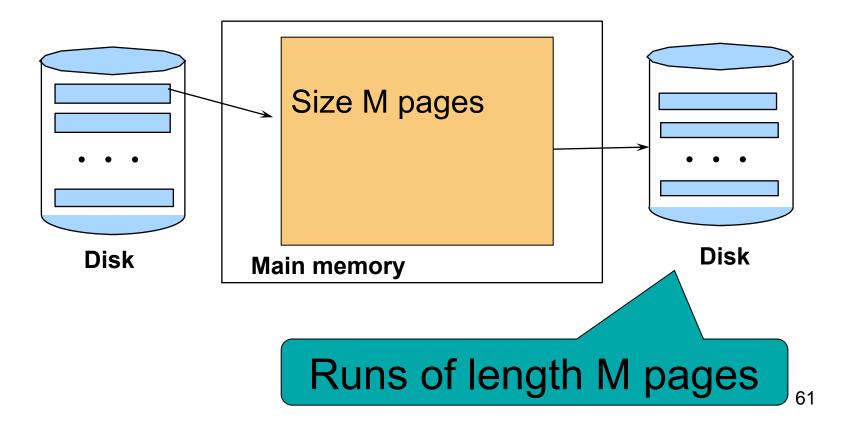
- How many I/Os ?
- Cost of partitioned hash join: 3B(R) + 3B(S)
- Hybrid join saves 2 I/Os for a t/k fraction of buckets
- Hybrid join saves 2t/k(B(R) + B(S)) I/Os
- Cost: (3-2t/k)(B(R) + B(S)) = (3-2M/B(S))(B(R) + B(S))

External Sorting

- Problem: Sort a file of size B with memory M
- Where we need this:
 - ORDER BY in SQL queries
 - Several physical operators
 - Bulk loading of B+-tree indexes.
- Will discuss only 2-pass sorting, for when $B < M^2$

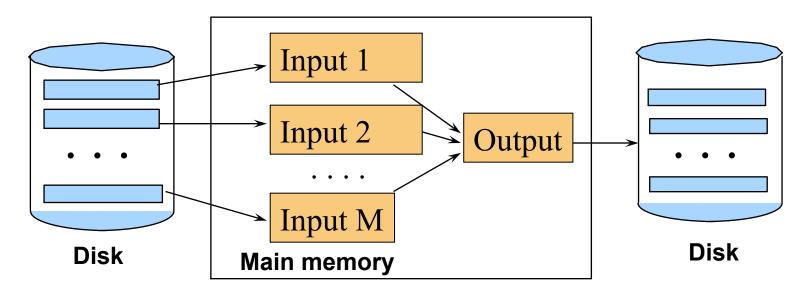
External Merge-Sort: Step 1

• Phase one: load M pages in memory, sort



External Merge-Sort: Step 2

- Merge M 1 runs into a new run
- Result: runs of length M (M 1) \approx M²



If $B \le M^2$ then we are done

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External Merge-Sort

- Cost:
 - Read+write+read = 3B(R)
 - Assumption: $B(R) \le M^2$
- Other considerations
 - In general, a lot of optimizations are possible

Duplicate elimination $\delta(R)$

- Trivial idea: sort first, then eliminate duplicates
- Step 1: sort chunks 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) <= M²

Grouping: $\gamma_{a, sum(b)}$ (R)

- Same as before: sort, then compute the sum(b) for each group of a's
- Total cost: 3B(R)
- Assumption: B(R) <= M²

 $\mathsf{Join} \ \mathsf{R} \bowtie \mathsf{S}$

- 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)
- Total cost: 5B(R)+5B(S)
- Assumption: $B(R) \le M^2$, $B(S) \le M^2$

 $\mathsf{Join} \ \mathsf{R} \bowtie \mathsf{S}$

• If $B(R) + B(S) \le M^2$

Or if we use a priority queue to create runs of length 2|M| (see paper)

- If the number of tuples in R matching those in S is small (or vice versa) we can compute the join during the merge phase
- Total cost: 3B(R)+3B(S)

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Review: Access Methods

• Heap file

- Scan tuples one at the time

Hash-based index

- Efficient selection on equality predicates
- Can also scan data entries in index

• Tree-based index

- Efficient selection on equality or range predicates
- Can also scan data entries in index

Index Based Selection

- Selection on equality: $\sigma_{a=v}(R)$
- V(R, a) = # of distinct values of attribute a
- Clustered index on a: cost B(R)/V(R,a)
- Unclustered index on a: cost T(R)/V(R,a)
- Note: we ignored the I/O cost for the index pages

Index Based Selection

• Example:

B(R) = 2000T(R) = 100,000 V(R, a) = 20

cost of $\sigma_{a=v}(R) = ?$

- Table scan (assuming R is clustered)
 - B(R) = 2,000 I/Os
- Index based selection
 - If index is clustered: B(R)/V(R,a) = 100 I/Os
 - If index is unclustered: T(R)/V(R,a) = 5,000 I/Os
- Lesson
 - Don't build unclustered indexes when V(R,a) is small !

Index Nested Loop Join

$\mathsf{R}\bowtie\mathsf{S}$

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

Summary of External Join Algorithms

- Block Nested Loop Join: B(R) + B(R)*B(S)/M
- Hybrid Hash Join: (3-2M/B(S))(B(R) + B(S)) Assuming t/k * B(S) >> k-t
- Sort-Merge Join: 3B(R)+3B(S)
 Assuming B(R)+B(S) <= M²
- Index Nested Loop Join: B(R) + T(R)B(S)/V(S,a) Assuming R is clustered and S has clustered index on a

Summary of Query Execution

- For each logical query plan
 - There exist many physical query plans
 - Each plan has a different cost
 - Cost depends on the data
- Additionally, for each query
 - There exist several logical plans
- Next lecture: query optimization
 - How to compute the cost of a complete plan?
 - How to pick a good query plan for a query?