# CSE 544 Principles of Database Management Systems 

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

## 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. $4^{\text {th }}$ 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 $\cup$, intersection $\cap$, difference -
- Selection $\sigma$
- Projection $\pi$
- Join $\bowtie$
- Duplicate elimination $\delta$
- Grouping and aggregation $\gamma$
- Sorting $\tau$
- Rename $\rho$


## Logical Query Plan



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

## 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
```



## 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

(On the fly)
$\pi$ sname
(On the fly) $\sigma_{\text {sscity }}=‘$ Seattle' $\wedge$ sstate='WA' $\wedge p n o=2$
(Nested loop)


Suppliers (File scan)

Supplies
(File scan)

## 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

(On the fly)
$\pi$ sname
(On the fly) $\sigma_{\text {sscity }}=' S e a t t l e ' ~ \wedge s s t a t e=' W A^{\prime} \wedge ~ p n o=2$
(Nested loop)


Suppliers (File scan)

Supplies
(File scan)

## 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

(On the fly)

(Scan: write to T1)

(Scan: write to T2)
$\sigma_{\text {sscity }}=$ 'Seattle' $\wedge$ sstate='WA'

Suppliers
(File scan)

Supplies
(File scan)

## Outline

- Steps involved in processing a query
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## 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: $T(R)+2 B(R)$


## One-pass Algorithms

Selection $\sigma(\mathrm{R})$, projection $\Pi(\mathrm{R})$

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



## Join Algorithms

- Logical operator:
- Product(pname, cname) $\bowtie$ Company(cname, city)
- 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 $R \bowtie S$
- $R$ is the outer relation, $S$ is the 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: $B(R)+T(R) B(S)$ 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


## Nested Loop Joins

- We can be much more clever
- How would you compute the join in the following cases ? What is the cost?
$-B(R)=1000, B(S)=2, M=4$
$-B(R)=1000, B(S)=3, M=4$
$-B(R)=1000, B(S)=6, M=4$


## Nested Loop Joins

- Block Nested Loop Join
- Group of ( $\mathrm{M}-2$ ) 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 $p r$ do
if "r and $s$ join" then output( $r, s$ )


## Nested Loop Joins



## Nested Loop Joins

- 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: $\mathrm{B}(\mathrm{S})+\mathrm{B}(\mathrm{S}) \mathrm{B}(\mathrm{R}) /(\mathrm{M}-2)$
- 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: $B(\delta(R))<=M$


## One-pass Algorithms

Grouping:
Product(name, department, quantity)
$\gamma_{\text {department, sum(quantity) }}$ (Product) $\rightarrow$ Answer(department, sum)
How can we compute this in main memory?

## One-pass Algorithms

- Grouping: $\gamma_{\text {department, sum(quantity) }}(R)$
- 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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## 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 ?
- Yes if $B(R) / M<=M$, i.e. $B(R)<=M^{2}$


## Hash Based Algorithms for $\delta$

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


## Hash Based Algorithms for $\gamma$

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


## Simple Hash Join

$R \bowtie S$

- 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


## Simple Hash Join

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

Original


## 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


## Partitioned (Grace) Hash Join

$R \bowtie S$

- 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 .



## Partitioned Hash Join

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



## Partitioned Hash Join

- Cost: 3B(R) + 3B(S)
- Assumption: $\min (B(R), B(S))<=M^{2}$


## Hybrid Hash Join Algorithm

- Assume we have extra memory available
- Partition S into k buckets
$t$ buckets $S_{1}, \ldots, S_{t}$ stay in memory
$k$-t buckets $S_{t+1}, \ldots, S_{k}$ 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:

$$
\left(\mathrm{R}_{\mathrm{t}+1}, \mathrm{~S}_{\mathrm{t}+1}\right),\left(\mathrm{R}_{\mathrm{t}+2}, \mathrm{~S}_{\mathrm{t}+2}\right), \ldots,\left(\mathrm{R}_{\mathrm{k}}, \mathrm{~S}_{\mathrm{k}}\right)
$$

## Hybrid Hash Join Algorithm

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

$$
\begin{aligned}
& k<=M \\
& t / k * B(S)<=M \\
& t / k * B(S)+k-t<=M
\end{aligned}
$$

- Assuming $\mathrm{t} / \mathrm{k}$ * $\mathrm{B}(\mathrm{S}) \gg \mathrm{k}-\mathrm{t}: \mathrm{t} / \mathrm{k}=\mathrm{M} / \mathrm{B}(\mathrm{S})$


## Hybrid Hash Join Algorithm

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


## 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 $\mathrm{M}-1$ runs into a new run
- Result: runs of length $M(M-1) \approx M^{2}$



## External Merge-Sort

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


## Two-Pass Algorithms Based on Sorting

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 - $\operatorname{cost} B(R)$
- Total cost: $3 B(R)$, Assumption: $B(R)<=M^{2}$


## Two-Pass Algorithms Based on Sorting

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

- Same as before: sort, then compute the sum(b) for each group of a's
- Total cost: 3B(R)
- Assumption: $B(R)<=M^{2}$


## Two-Pass Algorithms Based on Sorting

## Join $R \bowtie S$

- Start by sorting both $R$ and $S$ on the join attribute:
- Cost: $4 \mathrm{~B}(\mathrm{R})+4 \mathrm{~B}(\mathrm{~S})$ (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: $B(R)<=M^{2}, B(S)<=M^{2}$


## Two-Pass Algorithms Based on Sorting

## Join $R \bowtie S$

- If $B(R)+B(S)<=M^{2}$

Or if we use a priority queue to create runs of length $2|\mathrm{M}|$ (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: $\sigma_{a=v}(R)$
- $\mathrm{V}(\mathrm{R}, \mathrm{a})=$ \# 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:

$$
\begin{aligned}
& \mathrm{B}(\mathrm{R})=2000 \\
& \mathrm{~T}(\mathrm{R})=100,000 \\
& \mathrm{~V}(\mathrm{R}, \mathrm{a})=20
\end{aligned}
$$

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

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


## Index Nested Loop Join

$R \bowtie S$

- Assume $S$ has an index on the join attribute
- Iterate over R, for each tuple fetch corresponding tuple(s) from $S$
- 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 $B(R)+B(S)<=M^{2}$

- 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?

