



- 1. Introduction
- 2. Physical-Query-Plan Operators
- 3. Query Optimization

Reference:

Chapter 13-14, Database System Concepts, Sixth Edition Chapter 15, Database System, the Complete Book.





```
table course(course_id, title, dept_name);
table instructor(ID, name, dept_name, salary);
table teaches(ID, course_id, sec_id, semester, year);
SELECT name, title
FROM instructor
   NATURAL JOIN teaches
   NATURAL JOIN course
WHERE dept_name='Music';
```

- Use the knowledge we have learnt this week to discuss:
 - What are the possible ways to answer the above SQL query?
 - Which one is better?





```
SELECT name, title
FROM instructor
NATURAL JOIN teaches
NATURAL JOIN course
WHERE dept_name='Music';
```

Relational Algebra

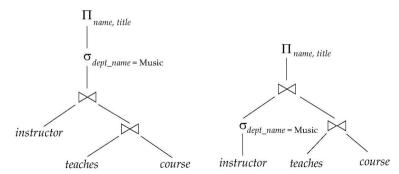
- $\Pi_{name,title}(\sigma_{dept_name=Music}(instructor \bowtie (teaches \bowtie course))$
- $\Pi_{name,title}(\sigma_{dept_name=Music}(instructor) \bowtie (teaches \bowtie course))$





- $\Pi_{name,title}(\sigma_{dept_name=Music}(instructor \bowtie (teaches \bowtie course))$
- $\Pi_{name,title}(\sigma_{dept_name=Music}(instructor) \bowtie (teaches \bowtie course))$

Query plan (Logic plan):



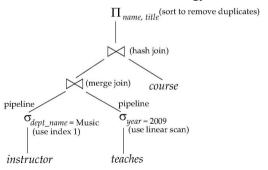




■ $\Pi_{name,title}((\sigma_{dept_name=Music}instructor \bowtie \sigma_{year=2009}teaches) \bowtie course)$

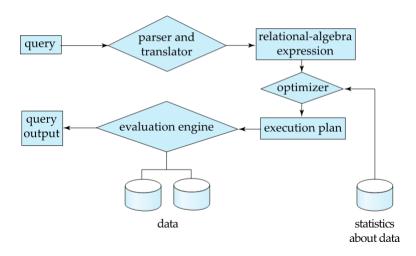
Physical/Evaluation plan

: an annotated expression on detailed evaluation strategy.



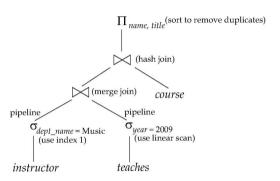












Cost estimation The cost of an operator is dependent on

- Metadata: The size of the underlying relation.
- Implementation: E.g., table scan, or using indexes (B+tree/hashing/ \cdots).
- reporting: Whether the result needs to be stored on disk.

In this course, we assume that we don't need to store the guery results on the disk.







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 - One-pass algorithms
 - Multi-pass algorithms
 - Index-based algorithms







External Memory Model

Denote by B the block size of the system. Let M be the size of the main memory. The cost of an algorithm is decided by the number of blocks (# of I/Os) transferred between the main memory and the disk.

Sometimes we use m=M/B pages as the size of the main memory to simplify the analysis. In this scenario, the allocation of the memory is in pages.





Physical-Query-Plan Operators

Iterators:

```
Open() {
   b := the first block of R:
   t := the first tuple of block b;
GetNext() {
   IF (t is past the last tuple on block b) {
        increment b to the next block;
        IF (there is no next block)
            RETURN NotFound:
        ELSE /* b is a new block */
            t := first tuple on block b:
    } /* now we are ready to return t and increment */
   oldt := t;
    increment t to the next tuple of b;
    RETURN oldt;
Close() {
```







Iterators:

- Combines several operations into one
- Avoids writing temporary files
- Many iterators may be active at one time







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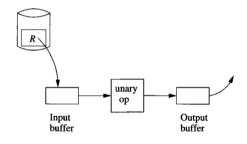
- Tuple-at-a-time, unary operations: selection and projection
- Full-relation, unary operations: the grouping operator and duplication deduction operator, conditioned that the relation can fit in to the main memory.
- Full-relation, binary operations: unions, intersection, difference, join, product, conditioned that one of the relation fit into the main memory.







■ Tuple-at-a-time, unary operations: selection and projection

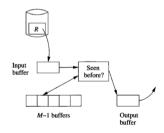






One-pass algorithms

- Tuple-at-a-time, unary operations: selection and projection
- Full-relation, unary operations: the grouping operator and duplication deduction operator, conditioned that the relation can fit in to the main memory.







- Tuple-at-a-time, unary operations: selection and projection
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Arrangement of the memory:

- M-B-1: Hold one relation
- B: buffer one page of the other relation
- 1: buffer for the output







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 - Multi-way merge sort
 - Join with both relations larger than the memory size
 - Index-based algorithms

Multi-way Merge Sort





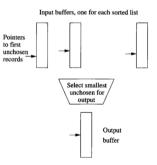
Sort-based algorithms:

- Duplication elimination
- Grouping and aggregation
- Union
- Intersection and difference





Atom operation: merge m-1 sorted list into one sorted list by allocating m-1 input buffer and 1 output buffer. The cost is linear to the size of the data.



Multi-way Merge Sort

Sort the data of n blocks on a machine with a memory of m blocks.

- Create runs (pass 0):
 - Each time load m blocks into memory, sort them and then output to create a run
- Merge pass i, $i = 1, 2, \cdots$: for every m-1 runs created by pass i-1
 - synchronize scan these runs to create a single run
 - one block used for output buffer until a pass creates only one run.

Complexity.

 $n \log_{m-1} \frac{n}{m} + n \approx n \log_m n \text{ I/Os}$

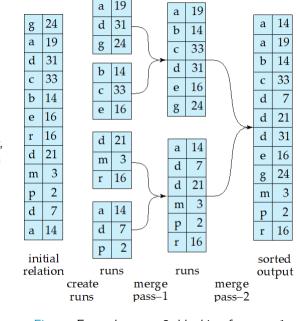


Figure: Example: m = 3, blocking factor = 1

Join



- Block nested-loop join
- Merge join
- Hash join





```
r \bowtie s: r is stored in n_r blocks. s is stored in n_s blocks
for each chunk of m-1 blocks of r
  read these blocks into the main memory buffers;
  organize their tuples into a search structure whose search key is the

→ common attributes of r and s:
  for each block bs of s
    load bs into the main memory
      for each tuple t of bs
         find the tuples of r in the buffer that join with t
        report the result
Complexity: \lceil n_r/(m-1) \rceil n_s + n_r \rfloor N_s.
```

Merge Join

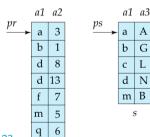


 $r \bowtie s$: r is stored in n_r blocks. s is stored in n_s blocks.

Join attribute: A.

Assume that r and s are both ordered A.

for each value x of attribute A shared by s and r (synchronize scan) block-nested-loop join on subrelations s[A=x] and r[A=x]



Complexity.

- Best case: $n_r + n_s$
- Worst case: block-nested-loop-join
- Consider the cost when
 - A is a key of r, or
 - all the attribute values are having the same frequency



Hashing: Partition a relation r to k buckets.

```
Consider a relation r with an attribute A and an integer k \in [1, m);
Hash function h_A() maps a tuple t to an integer in range [0,k) based on its value on A.
initialize k buckets using k empty buffers;
for each block b of relation r
  read block b into the (k+1)-th buffer:
  for each tuple t in b
  if the buffer for bucket h(t) has no room for t
    write the buffer to disk;
    initialize a new empty block in that buffer;
    copy t to the buffer for bucket h(t);
for each bucket do
  if buffer for this bucket is not empty
    write the buffer to disk:
  Denote by r_i the i-th bucket of r.
```





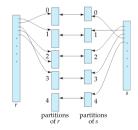
Hash Join

 $r \bowtie s$: r is stored in n_r blocks. s is stored in n_s blocks.

Join attribute: A.

Hash function $h_A()$ maps from a tuple to an integer range [0, k) based on A.

- 1. Partition r to k buckets using hash function h_A
- 2. Partition s to k buckets using hash function h_A
- 3. For each integer $i \in [0, k)$
 - Join on r_i and s_i (one pass/multi pass)







Goal is to count disk I/Os. But we First have to estimate sizes of intermediate results?

- \bigvee V(A,r) number of distinct values of attribute A in relation r.
- |r| number of tuples in relation r.

Consider relation r(A, B) with $n_r = 1000$ blocks and relation s(A, C) with $n_s = 500$ blocks. The memory had m = 101 pages.

- Nested-loop-join
- Multi-way merge join
- Hash join



Examples: Cost Estimation, Nested-Loop-Join

Consider relation r(A, B) with $n_r = 1000$ blocks and relation s(A, C) with $n_s = 500$ blocks. The memory had m = 101 pages.

Outer-loop: load chunk of 100 blocks of s to the main memory

5 chunks, 100 blocks each

- Inner-loop: scan *r* in 1000 blocks
- Total cost 5500 I/Os.

Switch the inner and outer loop:

- Outer-loop: load chunk of 100 blocks of *r* to the main memory 10 chunks, 100 blocks each
- Inner-loop: scan r in 500 blocks
- Total cost 6000 I/Os.

Conclusion: slight advantage in having the smaller relation on the outer loop.



Examples: Cost Estimation, Merge Join

Consider relation r(A, B) with $n_r = 1000$ blocks and relation s(A, C) with $n_s = 500$ blocks. The memory had m = 101 pages.

- The sorting cost of r: 4000 (two reads and two writes per block)
- The sorting cost of s: 2000 (two reads and two writes per block)
- Merge join: If *A* is the key of one relation, then the cost is 1500 blocks.
- Total cost 6500 I/Os.

Why linear time (when A is a key) merge join is not as good as quadratic time nested loop join in this case?



Examples: Cost Estimation, Hash Join

Consider relation r(A, B) with $n_r = 1000$ blocks and relation s(A, C) with $n_s = 500$ blocks. The memory had m = 101 pages. Let k = 100.

- The average size for each bucket is 10 blocks for relation r and 5 for s.
- Partition *n* and *s* (linear time): $1500 \times 2 = 3000 \text{ I/Os}$
- Since for each $i \in [0, k)$, the buckets of r and s can altogether fit in main memory, one-pass join: 1500 I/Os for loading.
- Total cost: 4500 I/Os.







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Application of Index in

- Selection
- Join

To avoid/reduce the number of table scan.

Index-based Algorithms: Cost Estimation Assumptions





We first have to estimate sizes of the intermediate results.

- V(A, r) number of distinct values of attribute A in relation r.
- |r| number of tuples in relation r.
- \blacksquare n_r number of blocks used to store relation r.

Important assumptions:

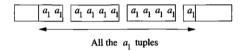
- Simple selection: All V(A, r) values are equally likely for attribute A; in other words, $|\sigma_{A=c}r| = |r|/V(A, r)$.
- Selection involving inequality $|\sigma_{A < c} r|$: Common assumption that 1/3 will meet the condition.
- Complex conditions AND: use decompositions.
 - $|\sigma_{A=c \text{ and } B < d}r| = |r|/3V(A, r).$





Index-based Algorithms: Selection

Sequential file:



The actual cost of $\sigma_{A=v}r$ can be slightly larger than $n_r/V(A,r)$.

- The index is not kept entirely in main memory, some disk I/Os are need to support the index lookup.
- Even when all the tuples with A = v might fit in b blocks, they could be spread over b+1 blocks because they don't start at the beginning of a block.
- The blocks can be not full, e.g., the B^+ tree's leaf nodes can be not full.

Unordered file

■ We assume that we need to visit |r|/V(A,r) blocks for answering $\sigma_{A=v}r$.







Assume that for relation r(A, B), $n_r = 1000$ and |r| = 20,000. Consider $\sigma_{A=0}r$. We ignore the cost accessing the index blocks in all cases.

r is seque	ntial on	A but	we do	not use	the index
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r is unordered on A but we use an index on A

ightharpoonup r is sequential on A, V(A,r)=100, use an index

ightharpoonup r is unordered on A, V(A,r)=10, use an index

 \blacksquare A is the key of r, use an index

1000 I/Os.

20,000 I/Os.

10 I/Os.

2000 I/Os.

1 I/O.





Index-based Algorithms: Nested-Loop Join

```
r(A, B) \bowtie s(A, C):
```

- ightharpoonup r has |r| tuples on n_r blocks, s is stored in n_s blocks
- s has an index on A;

for each tuple t of r

Get all tuples of s that can join with t using the index

Complexity: For each tuple t of r, an average of |s|/V(A,s) tuples should be retrieved, the cost is dependent on s

- If s is sequential, for each t the cost is $n_s/V(A,s)$ I/Os;
- Otherwise, the cost is |s|/V(A, s) I/Os.



Index-based Algorithms: Nested-Loop Join

Example: Consider r(A, B) with $n_r = 1000$ and s(A, C) with $n_s = 500$ while ten tuples of either relation fit in one block. Therefore, |r| = 10,000 and |s| = 5000. Assume that V(A, s) = 100. Suppose that s is ordered on A and there is a clustering index of s on a. Compute the cost of a is a.

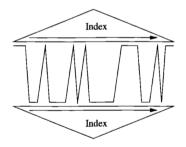
■ The number of I/Os in accessing s is $10000 \times 500/100 = 50000$ I/Os.

If r is considerably smaller than s then the index-based nested-loop join would be better than nested-loop join.



Index-based Algorithms: Merge Join

Both r and s are sequentially stored, each having a clustered index on A.









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Thank you for your attention!

Any questions?