Query Processing Part 1: A Simplified Cost Model and the Scan Operator

Abdu Alawini

University of Illinois at Urbana-Champaign

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Leaning Objectives

After this lecture, you should be able to:

- Explain the big picture of Query Processing and Optimization
- Discuss the difference between physical and logical operators
- Explain the basic model of costing for physical operators
- Reason about the different implementations of scan operators and their cost/memory requirements

Today's Lecture!

- Query Execution: The Big Picture
 - Basics of Costing
 - Classifications of Physical Operators
- Scan Operator
 - Table-scan, index-scan and sort-scan
 - Costs for Scans

RECAP: Codd's Logical Operations: The Relational Algebra

Six basic operations:

• Projection $\pi_{\alpha}(R)$

• Selection $\sigma_{\theta}(R)$

• (Rename) ρ_{α} (R)

• Union $R_1 \cup R_2$

• Difference $R_1 - R_2$

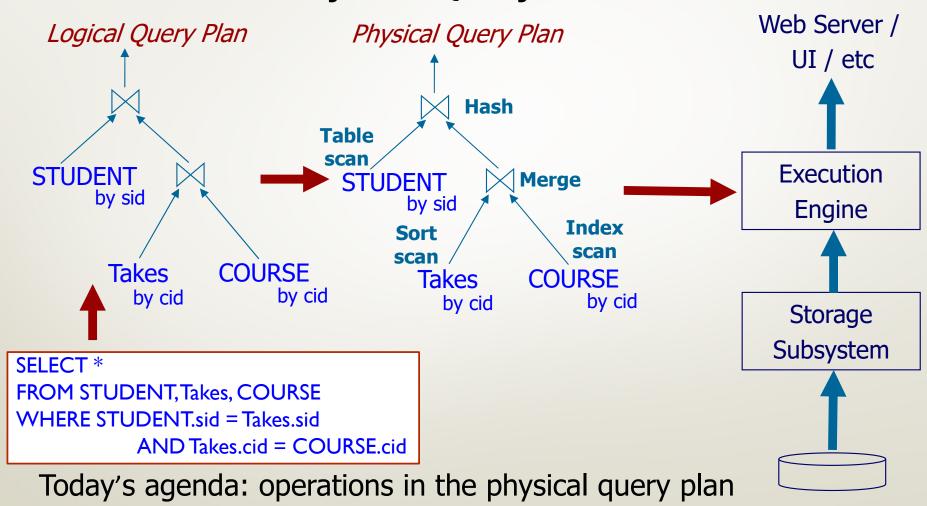
• Product $R_1 \times R_2$

And some other useful ones:

• Join $R_1 \bowtie_{\theta} R_2$

•Intersection $R_1 \cap R_2$

QP: The Big Picture: SQL → Logical Query Plan → Physical Query Plan



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/sort-merge/hash/index join
 - In other words, physical operators are particular implementations of relational algebra operations
 - Physical operators also pertain to none RA operations, such as "scanning" a table.

Physical operators and costs

Before we describe implementations, we need to examine the *cost* involved

- •Often, we have to make choices about which physical operators to use.
- •For this, we need to estimate the "cost" of each physical operator.

Cost Parameters (or "statistics")

Estimating the cost:

- Important in optimization (next topic: Qry Opt.)
 - Picking between plans
- Compute disk I/O cost only
 - Main memory operations are cheap
- We compute the cost to read the arguments of the operator
- We don't compute the cost to write the "final" result
 - Same for all types of operators

Cost Parameters (or "statistics")

- Cost parameters
 - M = number of blocks that fit in main memory
 - \bullet B(R) = number of blocks needed to hold R (best case # blocks)
 - $^{\bullet}$ T(R) = number of tuples in R (worst case # of blocks)
 - $^{\bullet}$ V(R,a) = number of distinct values of the attribute a

T(R) / (number of tuples that fits in a block) = B(R)

Here
$$T(R) = 4$$
; $B(R) = 2$

Clustered File



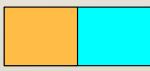


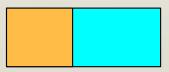


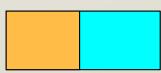


Unclustered File

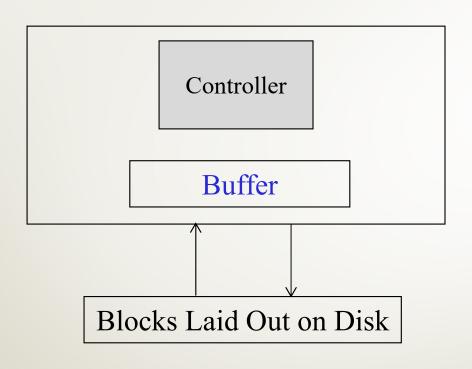








A Mental Model for Costing Operator Implementations



Temp computation space in main memory

We compute the cost to *read* the arguments of the operator We don't compute the cost to *write* the result

Think-Pair-Share

In general, which file organization would be more efficient in terms of disk I/Os

A: Clustered file

B: Unclustered file

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- Query Execution: The Big Picture
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- Nested-loop joins

The iterator model for implementing operators

Each (physical) operation is implemented by 3 functions:

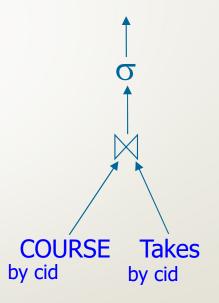
 Open: sets up the data structures and performs initializations

• GetNext: returns the the next tuple of the result.

• Close: ends the operations. Cleans up the data structures.

The iterator model for implementing operators

- Enables pipelining!
 - As opposed to: execute each operator in entirety, store its results on disk or in main memory
 - Many operators can be "active" simultaneously
- Pipelining: Not always possible (or meaningful):
 - E.g., "sort scan".
 - "Blocking" operators

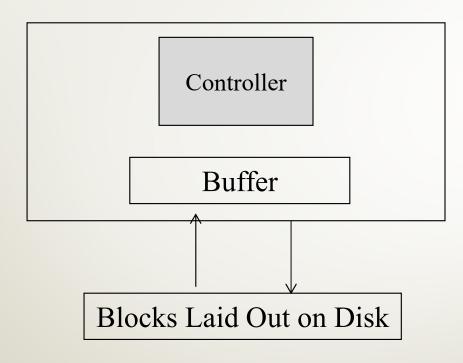


Overview of operator implementations: first classification

Operator algorithms mostly of one of these types:

- Sorting-based
- Hash-based
- Index-based

Another classification of operator implementations



One pass:

- reading the data only once from disk.
- •Typically, at least one of the arguments must fit in memory.

Two pass:

Data need not fit in memory but is not "too large".

Multipass:

No limit on data size.

How to think about operators

- Classification 1: Hash/Index/Sort
- Classification 2: One/Two/Many pass
- Metric 1: I/O Cost
- Metric 2: Buffer requirements (how large should M be)

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Getting started: Scanning Tables

- Read the entire contents of a relation R
 - or partial contents: all tuples that satisfy a criterion
- Table-scan: R is stored in some area of secondary storage (disk), in blocks.
 - These blocks are known to the system.
 - Can get all blocks one by one.
- Index-scan: if we have index to find the blocks.
 - "Scan the index, grab the relevant blocks"
 - This can be particularly useful for getting tuples that satisfy a predicate

Sorting while scanning tables

- May want to sort the tuples as we read them: "Sort-scan"
- •Why?
 - ORDER BY in query
 - Some RA operations are implemented using sort
 - Union, Intersection, ...
 - Future operations may be cheaper (e.g., GROUP BY)
- •How?
 - If indexed, then trivial
 - If fits in main memory, then table-scan or index-scan and sort in memory
 - If too large, "multiway merge sort" (see later)

Sorting with Scanning

First cut: relation is small

- Have M main memory blocks available for use
- Assumption: $B(R) \le M$
- Method:
 - Read all B(R) blocks, sort, write
- Cost: B(R)

Sorting with Scanning

Second cut: Relation is large

- Have M main memory blocks available for use
- How would you do this?

Sorting with Scanning

Second cut: Two pass "multi-way merge sort"

- Have M main memory blocks available for use
- Step 1:
 - Read M blocks at a time, sort, write
 - Result: have runs of length M on disk
- Step 2:
 - Merge M-1 runs at a time, construct output, write to disk
 - Result: have runs of length $M(M-1) \approx M^2$

Example

M = 3, each block holds two values

Data on disk: [3,9] [8,1] [7,6] [9,5] [2,0] [3,8]

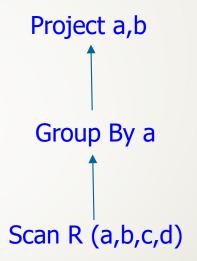
Example, cont.

```
M = 3, each block holds two values
Data on disk: [3,9] [8,1] [7,6] [9,5] [2,0] [3,8]
Step 1: Sort Runs
[3,9] [8,1] [7,6] => [1,3] [6,7] [8,9]
[9,5] [2,0] [3,8] => [0,2] [3,5] [8,9]
Step 2: Merge Sorted Runs
[1,3][0,2][\_,\_] => [1,3][0,2][0,1] => output [0,1]
[1,3][0,2][\_,\_] => [1,3][0,2][2,3] => \text{output } [2,3]
[6,7][3,5][\_,\_] => [6,7][3,5][3,5] => \text{output } [3,5]
[6,7][8,9][\_,\_] => [6,7][8,9][6,7] => output [6,7]
[8,9][8,9][, ] => [8,9][8,9][8,8] => output [8,8]
```

Think-Pair-Share

Given this query plan →

B(R)=20; M = 22; R is sorted on b which of the following implementation of table scan would be used?



A: Sort-Scan

B: Table-Scan

C: Multiway merge sort

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Multi-way merge sort cost

Cost: 3B(R),

Assumption: $B(R) < M^2$; why?

 $B(R) / M \le M-1 \Leftrightarrow B(R) \le M(M-1) < M^2$

Cost of the Scan Operator

- •Table scan: B(R); Sort-scan: 3B(R) [assuming B(R) <= M(M-1)]
- •Index scan: B(R) + #blocks of the index

 \approx B(R); Sort-scan: B(R)

(assuming index on sorting key)

Cost of the Scan Operator (Cont.)

- •Unclustered relation:
 - we have assumed so far that all tuples of R are "clustered", i.e., stored in ~ B blocks.
 - If tuples of R are interspersed with tuples of other relations, then cost:
 - •scan: T(R); sort: T(R) + 2B(R)

Now Generalizing...

Done with costing of scan operators

- Next lecture we will generalize to other operators
 - A lot more complex than scan
 - Many different classifications/tradeoffs

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