

Advanced Database Systems

Spring 2024

Lecture #19:

Query Evaluation: Processing Models

R&G: Chapter 14

PROCESSING MODEL

Processing model defines how the DBMS executes a query plan

Different trade-offs for different workloads

Three main approaches:

Iterator model

Vectorised (batch) model

Materialisation model

ITERATOR MODEL

Each query plan operator implements three functions:

open() - initialise the operator's internal state

next() - return either the next result tuple or a null marker if there are no more tuples

close() - clean up all allocated resources

Each operator instance maintains an internal state

Any operator can be input to any other (composability)

Since they all implement the same interface

Also called Volcano or Pipeline Model

Goetz Graefe. Volcano – An Extensible and Parallel Query Evaluation System. IEEE TKDE 1994

ITERATOR MODEL

Top-down plan processing

The whole plan is initially reset by calling open() on the root operator

The open() call is forwarded through the plan by the operators themselves

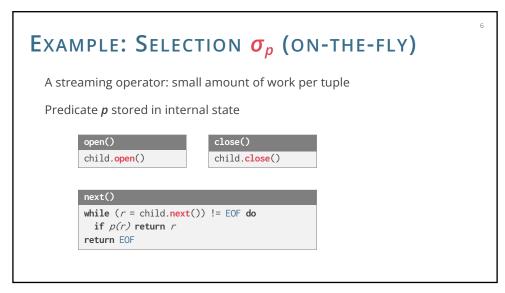
Control returns to the query processor

Operators forward the **next()** request as needed. As soon as the next result record is produced, control returns to the query processor again

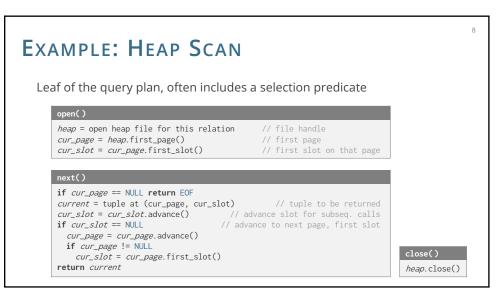
The root is requested to produce its **next()** result record

Used in almost every DBMS

TERATOR MODEL Query processor uses the following routine to evaluate a query plan Function eval(q) q.open() r = q.next() while r!= EOF do /* deliver record r (print, ship to DB client) */ emit(r) r = q.next() /* resource deallocation now */ q.close() Output control (e.g., LIMIT) works easily with this model



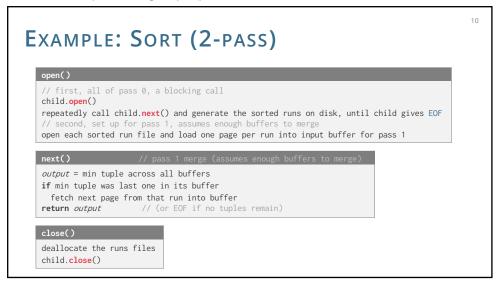
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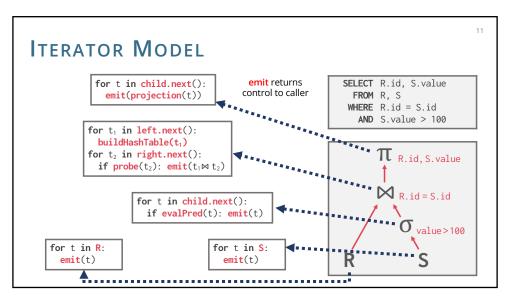


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EXAMPLE: NESTED LOOPS JOIN
   Volcano-style implementation of nested loops join R \bowtie_{p} S
          open()
                                              next()
           left_child.open()
                                              while r != EOF do
           right_child.open()
                                                while (s = right_child.next()) != EOF do
           r = left_child.next()
                                                 if p(r,s) return \langle r,s \rangle
                                                /* reset inner join input */
                                                right_child.close()
          close()
                                                right_child.open()
                                                r = left_child.next()
           left_child.close()
           right_child.close()
                                               return EOF
```

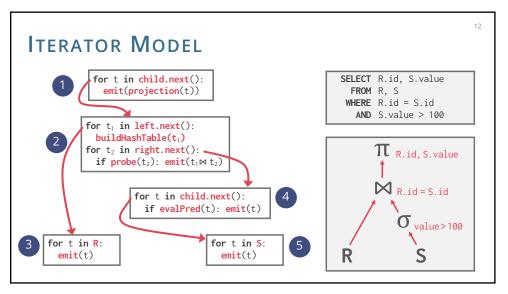
Sort is a blocking operation since we don't know the tuple that should be passed to its parent,

hence we complete sorting in open() function





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ITERATOR MODEL Allows for tuple pipelining The DBMS process a tuple through as many operators as possible before having to retrieve the next tuple Reduces memory requirements and response time since each chunk of input is propagated to the output immediately Some operators will block until children emit all of their tuples E.g., sorting, hash join, grouping and duplicate elimination over unsorted input, subqueries The data is typically buffered ("materialised") on disk

ITERATOR MODEL

+ Nice & simple interface

- + Allows for **easy** combination of operators
- Next called for **every single** tuple & operator
- Virtual call via function pointer
 Degrades branch prediction of modern CPUs
- Poor code locality and complex bookkeeping
 Each operator keeps state to know where to resume

VECTORISATION MODEL

Like Iterator Model, each operator implements a **next()** function

Each operator emits a batch of tuples instead of a single tuple

The operator's internal loop processes multiple tuples at a time

The size of the batch can vary based on hardware and query properties

Ideal for OLAP queries

Greatly reduces the number of invocations per operator

Operators can use vectorised (SIMD) instructions to process batches of tuples

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16 VECTORISATION MODEL for t in child.output(): SELECT R.id. S.value out.add(projection(t)) FROM R. S if |out| > n: emit(out) WHERE R.id = S.idAND S.value > 100 for t₁ in left.output(): buildHashTable(t₁) π R.id, S. value for t₂ in right.output(): if probe(t_2): out.add($t_1 \bowtie t_2$) if |out| > n: emit(out) \bowtie R.id = S.id 4 for t in child.output(): if evalPred(t): out.add(t) **O** value > 100 if |out| > n: emit(out) **out** = { } S for t in R: for t in S: out.add(t) out.add(t) if |out| > n: emit(out) if |out| > n: emit(out)

MATERIALISATION MODEL

Each operator processes its input all at once and then emits its output

The operator "materialises" its output as a single result

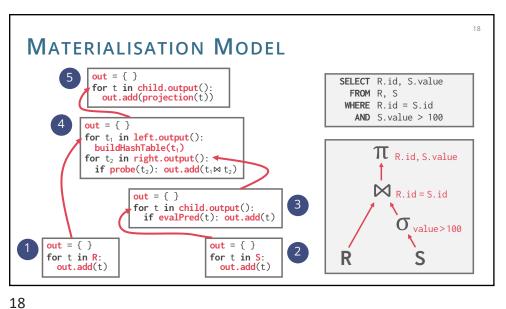
Bottom-up plan processing

Data not pulled by operators but **pushed** towards them Leads to better code and data locality

Better for OLTP workloads

OLTP queries typically only access a small number of tuples at a time Not good for OLAP queries with large intermediate results

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PROCESSING MODELS: SUMMARY

Iterator / Volcano

Direction: Top-Down Emits: Single Tuple Target: General Purpose

Vectorised

Direction: Top-Down Emits: Tuple Batch Target: OLAP

Materialisation

Direction: Bottom-Up Emits: Entire Tuple Set 19

Target: OLTP