# **DB** 2

13 - Plan Evaluation

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Torsten Grust Universität Tübingen, Germany

# 1 Evaluating Query Plan Trees



The evaluation of a (complex) query plan requires a coordinated execution of the plan's operators:

- Is data **pushed** from the leaves (e.g., Seq Scan, Index Scan) towards the plan root?
- Or does an operator **pull** the intermediate results from its upstream child operators?
- What kind of data flows across the plan's edges? Entire tables or columns? Single rows?
- Does the plan execute in one shot or can we **demand** the "next result row" when we are ready to consume it?
  - Can operators remember/resume from their current state?

# Query $Q_{12}$ and its (Moderately Complex) Plan



#### • **Q**<sub>12</sub>:

```
SELECT o.a, COUNT(*) AS "#"
FROM one AS o, many AS m
WHERE o.a = m.a
GROUP BY o.a
ORDER BY o.a DESC
```

• Plan operators:

```
1 Seq Scan on many (outer of 3)
2 Seq Scan on one (inner of 3)
3 Nested Loop (Join Filter: o.a = m.a)
4 HashAggregate (Group Key: o.a)
5 Sort (Sort Key: o.a DESC)
--- = direction of data flow
1 ... 5 = evaluation order
```

many one	1 3 3	b A B C	a 1 3	b A B C	b a c c		a 3 1	# 2 1	
			≟,		<b>]</b> —	 	<b>5</b> —	L	
	a 1 2 3	b а b с				1 3	1 2		



MonetDB generates MAL programs that evaluate operators following a post-order traversal<sup>1</sup> of the query plan tree.

- Leaf nodes evaluated first, downstream nodes consume BATs generated by child nodes. Root operator evaluated last.
- Each operator consumes entire BATs, generates and **fully materializes** its result BAT(s) [cf. previous slide].
  - Tight code loops process entire columns. Instruction
     and data locality, predictable memory access.
  - $\circ$  Size of intermediate results may exceed available RAM  $\Rightarrow$  OS-level paging and thus disk I/O.

 $<sup>^{1}</sup>$  Recall: data-flow dependency analysis enables the  $\parallel$  evaluation of  $\blacksquare$  and  $\blacksquare$ .

# Data Dependencies in MAL Program for $Q_{12}$

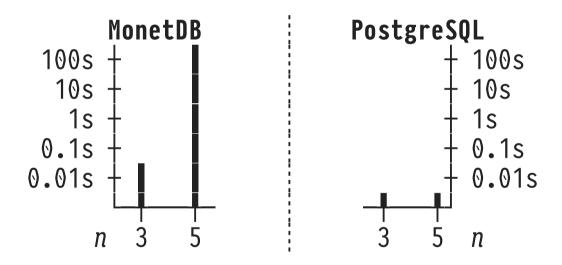


```
one :bat[:oid] := sql.tid(sql, "sys", "one");
one_a0 :bat[:int] := sql.bind(sql, "sys", "one", "a", 0:int);
                                                                                           1 Scan one.a
one a :bat[:int] := algebra.projection(one, one_a0);
                                                                                               1 / 2
       :bat[:oid] := sql.tid(sql| "sys", "many");
many_a0:bat[:int] := sql.bind(sq|, "sys", "many", "a", 0:int);
                                                                                           2 Scan manv.a
many_a :bat[:int] := algebra.pro|ection(many, many_a0);
(left, right) := algebra.join(one_a, many_a, nil:bat, nil:bat, false, nil:lng);
                                                                                           3 Equi-Join
joined one a:bat[:int] := algebra.projection(left, one a);
(grouped_one_a, group_keys, group_sizes) := group.groupdone(joined_one_a);
keys a:bat[:int] := algebra.projection(group keys, joined one a);
                                                                                           4 Group + Agg
count|:bat[:lng] := aggr.subcount(grouped_one_a, grouped_one_a, group_keys, false);
(sorted a, oidx, gidx) := algebra.sort(keys a, true, true, false);
            :bat[:int] := algebra.projection(oidx, keys_a);
                                                                                           5 Sort
result a
result_count:bat[:lng] := algebra.projection(oidx, count);
```

# 2 | Materialization vs. Demand-Driven Pipelining

Consider  $Q_{13}$ , returning the single value 42:

```
SELECT 42 AS fortytwo FROM hundred AS h_1, ..., hundred AS h_n -- A 100^n rows LIMIT 1
```



# Volcano-Style Demand-Driven Pipelining





#### PostgreSQL implements the Volcano Iterator Model:

- Operator **demands** its subplan to produce the next row (i.e., the plan root drives the query evaluation).
- Operator delivers results **one row at a time**, avoids intermediate result materialization (if possible !):
  - ∘ Reduces query *response* time (first row delivered immediately, do not wait until result is complete). ₺
  - Reduces memory requirements (pass data row-by-row, not table at a time).



Volcano-style **demand-driven** pipelining bears some resemblance with **call-by-need** evaluation of (functional) programming languages:

- If function  $f(e_1,e_2)$  does not (always) need the value of expression  $e_2$ , then f(42,1/0) may evaluate just fine.
- With the demand-driven evaluation in Haskell<sup>2</sup>, consider:

```
sum [x/0 \mid x <- [1..10], x > 42] \rightarrow 0.0 length [x/0 \mid x <- [1..10]] \rightarrow 10 take 1 [(x,y) \mid x <- [1..], y <- [1..]] <math>\rightarrow [(1,1)] \leftarrow Q_{13}
```

<sup>&</sup>lt;sup>2</sup> Haskell is a *lazily* evaluated functional programming language, see http://haskell.org.

#### Query Response vs. Evaluation Time



In PostgreSQL's EXPLAIN output, query response (first row) and evaluation time (all rows) are distinguished:

- Both times may...
  - ... differ substantially (pipelined evaluation),
  - coincide (blocking operators—e.g., Sort—evaluate in full first, then deliver all rows from intermediate result buffer).



In Volcano-style demand-driven query evaluation, operators implement a simple API of three main methods:

- open(): Initialize operator and its internal state, forward open() request to upstream subplans as well.
- 2. next(): If required, forward next() upstream to request
  more input rows. Then deliver next output row (or ¼ if
  result complete).
- 3. close(): Release operator-internal state, forward close() request to upstream subplans as well.

Volcano-style call protocol: (open() next()\* close())\*.

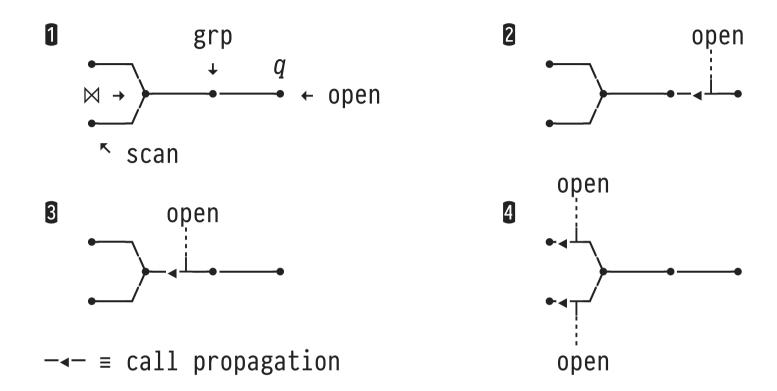


Use the Volcano iterator model API to fully evaluate a query. Operator q denotes the root of the query plan:

- To retrieve next result row only, simply call next(q).
- May/must use close(q) to cancel query evaluation midway.

# Volcano Iterator Model: Forwarding open()/close()





• Each operator instance (\*) allocates and releases its own copy of state that is kept between method invocations.



Implement open() and close() for the Nested Loop Join
operator:

```
NLJ.close(outer,inner,0):
    close(outer);
    close(inner);
```

### Pipelined Nested Loop Join (NLJ, cont'd)

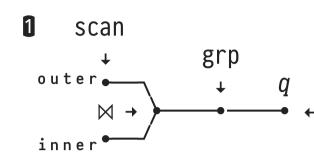


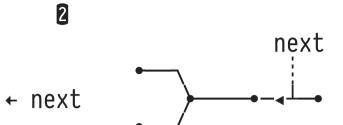
```
NLJ.next(outer,inner,0):
  forever
     if needNewOuter
         o ← next(outer);
                                    } o: current outer row
                                      no more outer rows

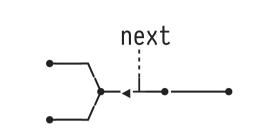
⇒ join complete
         if o = {}^{\mathsf{N}}
         L return ¼;
         needNewOuter ← false;
         close(inner);
                                      reset/rescan
inner input
         open(inner);
     i ← next(inner);
                                    } i: current inner row
                                      no more inner rows,
                                      next time: read new outer
       l needNewOuter ← true;
     else if o \theta i
                                    } join condition satisfied?
               return <0,i>;
                                    } return single joined pair
```

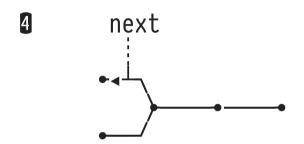
## Volcano Iterator Model: Evaluating a NLJ Plan

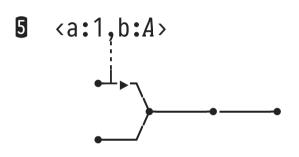


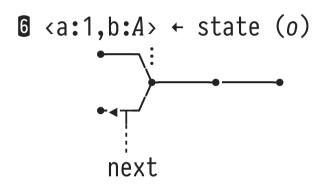


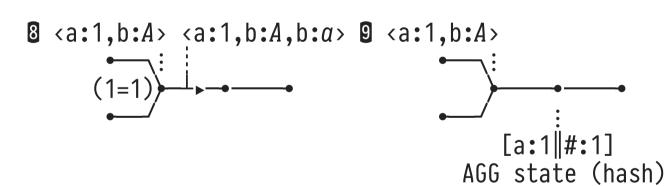








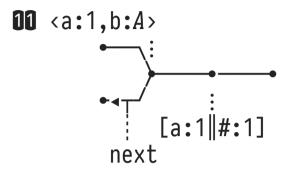


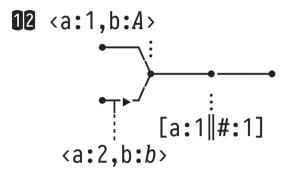


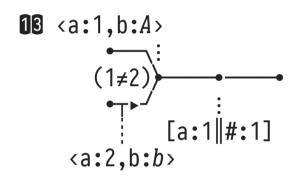
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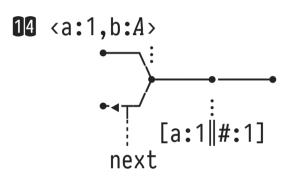
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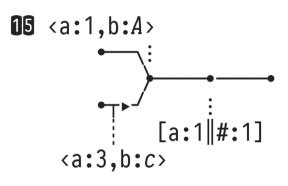


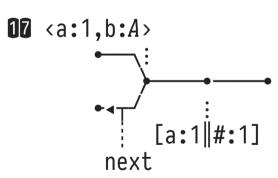






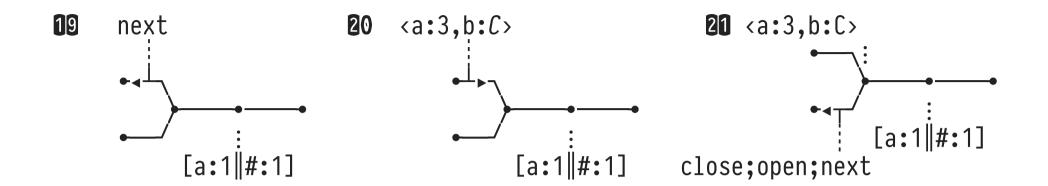






#### Volcano Iterator Model: Evaluating a NLJ Plan





**Quiz/Exercise:** Think about how to implement the following plan operators in the Volcano iterator model:

- Seq Scan (with Filter condition),
- Limit (given a row limit n),
- GroupAggregate (over input sorted by the Group Key),
- Append (SQL: UNION ALL).



Via cursors, the SQL standard exposes the Volcano-style open/next/close API at the level of (Embedded) SQL:

```
-- Generate query plan, no evaluation yet

1 DECLARE cursor [ SCROLL ] CURSOR FOR query
-- 

-- cursor can move backwards

-- Evaluate plan to deliver the next/prior row (n rows)

2 FETCH [ NEXT | [ FORWARD | BACKWARD ] n ] FROM cursor

-- Release plan/intermediate buffers

3 CLOSE cursor
```

• Statements need to be issued within an SQL transaction.



- Effectively, multiple operators are active at one time.
  - Aggregate intermediate state (memory) may be large.
  - Method call forwarding incurs function call overhead.
  - Frequent switches between code blocks due to row-by-row processing, CPU instruction cache misses are likely.
- ¶ Modern RDBMSs (e.g., VectorWise<sup>3</sup>, Umbra) seek middle ground between full materialization and pipelining:
  - Build demand-driven pipeline between operators, but...
  - ... pass vectors of rows—typically the size of the CPU's data cache—between operators.

<sup>&</sup>lt;sup>3</sup> See MonetDB/X100—A DBMS In The CPU Cache and MonetDB/X100: Hyper-Pipelining Query Execution.