DB 2

07 - Expression Evaluation

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1 Q_6 — Expression Evaluation

For a large class of queries, the **CPU effort to evaluate** (complex) expressions may easily match the time spent for I/O and data access:

Iterate over rows t, access required fields (here: t.a,
t.c), evaluate (multiple) expressions per row,
construct resulting row.

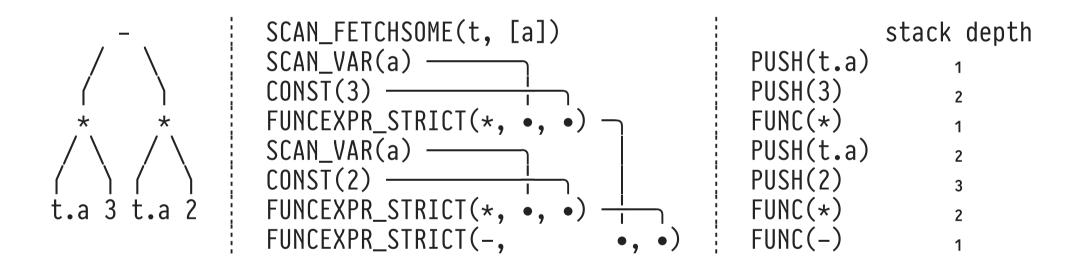
Using EXPLAIN on Q₆



• Expressions have been parenthesized, simplified, and annotated with type casts as required by SQL semantics.

Internal Representations of t.a * 3 - t.a * 2

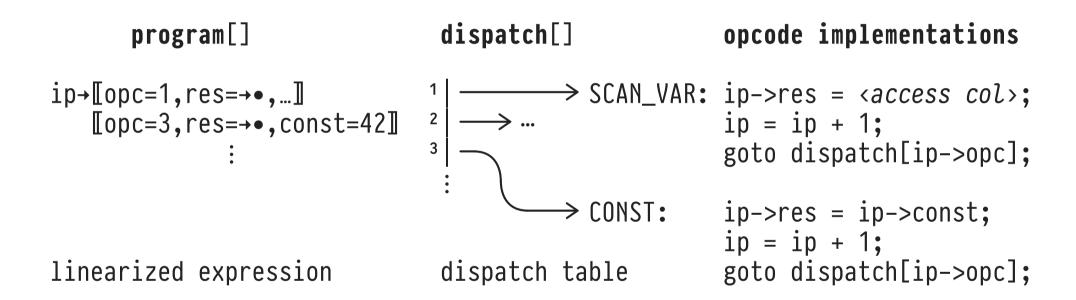
 DBMSs—just like interpreters and compilers—transform expressions into internal representations that facilitate simplification and evaluation:



 Postorder traversal of expression tree to obtain a linearized "program". Arg slots (●) or stack push/pop.



PostgreSQL implements a **threaded interpreter** over linearized expressions (middle column of previous slide):



- Note: ip: instruction pointer, opc: operation code.
- Relies on support for computed goto (e.g., common in C).

Expression Interpretation Overhead



Overhead of expression interpretation has been found to be **massive** in DBMS (cf. the threaded interpretation vs. machine code for t.a * 2).

- Field access and interpretation in hot query code path, rediscovers same row structure and follows same opcode pointers for every row processed. Wasteful.
- Invest in just-in-time (JIT) compilation of expression program into machine code once, benefit for all subsequent rows.
 - NB. LLVM-based support for JIT compilation of expressions has been added to PostgreSQL since v11.

JIT: Turn Run-time into Compile-time Decisions



- PostgreSQL's interpreter is very generic, prepared to handle corner cases, exceptions, and extensions. Leads to branch-heavy routines in the interpreter's hot code path.
- Expression compilation creates query-specific code:

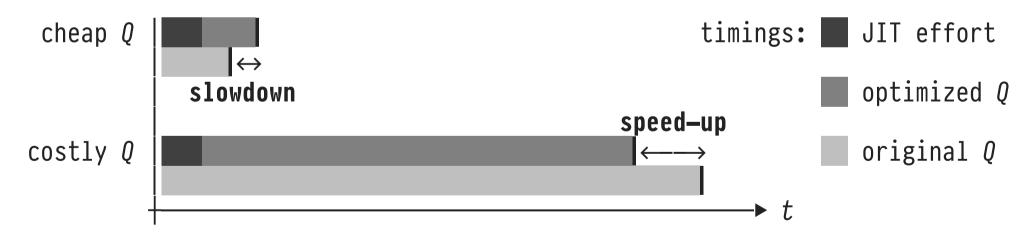
- Access field a of a row.
- Interpreter follows same code path for all rows, possibly millions of times.
- Generate specific "a access code" along the \ path.

Expression Compilation: When (Not) to JIT?



- JITing involves code generation, optimization, IR emission (LLVM bitcode), and translation to native code.
- JIT effort adds to the query planning time. Be careful not to penalize queries Q that are cheap to begin with:

Assume runtime of query Q reduced by 20% due to JIT compilation:



2 | Q₆ — Column-Based Expression Evaluation



MonetDB compiles expressions into sequences of MAL operations. Like data processing, expression evaluation is column-oriented (as opposed to row-by-row).

 We will find that this vector-based evaluation mode fits modern CPU architecture particularly well.



```
sql > EXPLAIN SELECT t.a * 3 - t.a * 2 AS a,
            ceil(t.c / log10(2)) AS bits
             FROM ternary AS t;
ternary :bat[:oid] := sql.tid(sql, "sys", "ternary");
        :bat[:dbl] := sql.bind(sql, "sys", "ternary", "c", 0:int);
C0
                   := algebra.projection(ternary, c0);
C
        :bat[:db1] := batcalc./(c, 0.3010299956639812:db1);
e1
       :bat[:dbl] := batmath.ceil(e1);
e2
                                                       result column 'bits'
        :bat[:int] := sql.bind(sql, "sys", "ternary", "a", 0:int);
a0
                   := algebra.projection(ternary, a0);
а
        :bat[:lng] := batcalc.lng(a);
e3
                                                           cast to type lng
        :bat[:lng] := batcalc.*(e3, 3:bte);
e4
        :bat[:lng] := batcalc.*(e3, 2:bte);
e5
        :bat[:lng] := batcalc.-(e4, e5);
e6
                                                          result column 'a'
```

 MAL ops batcalc.⊗ accept two BATs or one BAT + one scalar (like 2:bte, 3:bte, 0.301…:dbl = log10(2)).



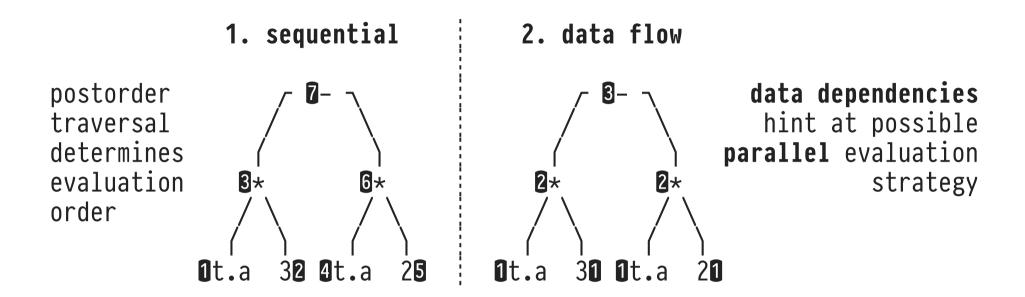
Operators batcalc.⊗ merge the tails of two synchronized BATs using binary operator ⊗, yields a new BAT:

X		<pre>batcalc(x,y)</pre>				у	
head	tail		head	tail		head	tail
0@0 1@0 2@0 3@0	X ₀ X ₁ nil X ₃	+ + + + +	000 100 200 300	X ₀ -y ₀ X ₁ -y ₁ nil X ₃ -y ₃	+ + +	000 100 200 300	y ₀ y ₁ y ₂ y ₃

• batcalc. \otimes contains checks for arithmetic exceptions (overflow, divide by 0). Also: nil $\otimes x = x \otimes \text{nil} = \text{nil}$.

MAL: Sequential Execution vs. Data Flow





- 1. Order of assignment to temporary result BATs e_i follows postorder traversal of expression tree.
- 2. Spawn CPU threads to evaluate data-independent subexpressions in // (see MonetDB's dataflow optimizer).

batcalc.⊗: Column-Based Operator Implementations (1)



MonetDB supplies type- and ⊗-specific implementations of MAL operations (code generation via C preprocessor macros):

```
/* batcalc.-(left:bat[:lng], right:bat[:lng]):bat[:lng] */
int i, j, k;
int nils = 0;
for (i = start, j = start*1, k = start; k < end; i += 1, j += 1, k += 1) {
 /* nil checking */
  if (is_lng_nil(left[i]) || is_lng_nil(right[j])) {
    result[k] = lng_nil;
   nils++;
 } else {
   /* omitted: overflow checking (abort on error or emit nil) */
   result[k] = left[i] - right[j];
```

batcalc.⊗: Column-Based Operator Implementations (2)



MonetDB supplies type- and ⊗-specific implementations of MAL operations (code generation via C preprocessor macros):

```
/* batcalc.-(left:bat[:lng], right:lng):bat[:lng] */
int i, j, k;
int nils = 0;
for (i = start, j = start*0, k = start; k < end; i += 1, j += 0, k += 1) {
 /* nil checking */
  if (is_lng_nil(left[i]) || is_lng_nil(right[j])) {
    result[k] = lng_nil;
   nils++;
 } else {
   /* omitted: overflow checking (abort on error or emit nil) */
   result[k] = left[i] - right[j];
```

3 | Column-Based Operators vs. Expression Interpretation

Expression evaluation through column-based operator and row-wise interpretation compared:

Column-Based (MonetDB)	Row-Wise (PostgreSQL)
zero degrees of freedom	variable-width rows w/ fields of various types
instruction locality	computed goto, long code paths
optimizable tight loops	complex control flow, code in many functions
optimizable tight loopsloop pipelining	• unpredictable branches
blocking	
 loop unrolling 	
data parallelism	focus on single row
data parallelism full materialization	row-by-row result generation

- Compilers optimize tight code loops inside MAL operators.
- CPUs offer wide registers and instructions to exploit data //ism (SIMD: single instruction, multiple data).

Compiling Tight Loops (cf. MAL Operators)



Inspect Intel® x86 code generated by LLVM's C compiler clang for MonetDB's routine BATcalcsub (batcalc.-), simplified:

```
#define SIZE 1024

void BATcalcsub(int *left, int *right, int *result)
{
   int i, j, k;
   for (i = j = k = 0; k < SIZE; i += 1, j += 1, k += 1) {
      result[k] = left[i] - right[j];
   }
}</pre>
```

• Arrays left, right/result represent input/output BATs.

Assembly Code for Simple Tight Loop



Uses clang (options -02 -fno-vectorize -fno-unroll-loops).

• Register assignment:

left: //rdi, right: //rsi, result: //rdx, i/j/k: //rax

• NB. One loop exit test per array element computed.

(Explicit) Loop Unrolling



- Manually perform loop unrolling to
 - 1. improve the ratio (useful work) / (loop exit test),
 - 2. expose independent work that may be executed in #:

```
void BATcalcsub(int *left, int *right, int *result)
{
  int i, j, k;

  for (i = j = k = 0; k < SIZE; i += 4, j += 4, k += 4) {
     result[k ] = left[i ] - right[j ];
     result[k+1] = left[i+1] - right[j+1];
     result[k+2] = left[i+2] - right[j+2];
     result[k+3] = left[i+3] - right[j+3];
}
</pre>
```

• NB. Needs code to handle the case SIZE mod $4 \neq 0$.



Compiler clang (options -02 -fno-vectorize -funroll-loops) unrolls four loop iterations (easy for CPU to //ize):

```
BATcalcsub:
                               # i/i/k \leftarrow_{32} 0
 xorl %eax, %eax
loop:
 movl (%rdi,%rax,4), %ecx
                            # %ecx ←32 left[i]
                           # %ecx ←32 %ecx -32 right[j]
  subl (%rsi,%rax,4), %ecx
 movl %ecx, (%rdx,%rax,4)
                           # result[k] +32 %ecx
  movl 4(%rdi,%rax,4), %ecx
                              # %ecx ←32 left[i+1]
  subl 4(%rsi,%rax,4), %ecx
                                # %ecx ←32 %ecx -32 right[j+1]
  movl %ecx, 4(%rdx,%rax,4)
                                # result[k+1] ←32 %ecx
  movl 8(%rdi,%rax,4), %ecx
                                # :
  subl 8(%rsi,%rax,4), %ecx
  movl %ecx, 8(%rdx,%rax,4)
  movl 12(%rdi,%rax,4), %ecx
  subl 12(%rsi,%rax,4), %ecx
  movl %ecx, 12(%rdx,%rax,4)
  addq $4, %rax
                                   1024 / 4 = 256 loop iterations
  cmpg $1024, %rax
                                # exit if \frac{1024}{1}
 jne loop
  reta
```

Data-Parallelism Through SIMD



result	[]	le	ft[]	r	ight[
0] 1] 2]	← 1:	2 8	 	⁻ 128	
3 🛛] 4 🗓] 5 🗍 } 6 🗒] 7 🗒]	←1:	2 8		⁻ 128	

- Read/compute/write four array elements (of width 4 × 32 bits = 128 bits) at a time in data-parallel fashion.
- Relies on SIMD register and instructions (e.g., Intel® SSE registers %xmm; and instruction move double quad word)

• A Requires care if

8 🛮

- arrays result[] and left[]/right[] overlap in memory,
- ∘ residual array elements (see □) are to be processed.

Data-Parallelism Through SIMD (Array Overlap)



Compiler clang (options -02 -fvectorize) uses SIMD registers and instructions.

- Extra prelude code checks for **array overlap**. If so, jumps to non-vectorized (yet unrolled) version of code.
- Declare function arguments via restrict to inform C compiler that arrays won't overlap:

```
void BATcalcsub(int *restrict left, ..., int *restrict result)
{
  int i, j, k;
  :
}
```

Data-Parallelism Through SIMD (Main Loop)



Process 16 elements per iteration (SIMD + 2 loops unrolled, assumes no overlap of arrays result[] and left[]/right[]):

```
xorl %eax, %eax
                                            4 \times 32 bits = 128 bits wide
loop:
 movdqu (%rdi,%rax,4), %xmm0
                                     # %xmm0 ←128 left[i+0...i+3]
 movdqu 16(%rdi,%rax,4), %xmm1
                                     # xmm1 \leftarrow_{128} left[i+4...i+7]
 movdqu (%rsi,%rax,4), %xmm2
                                      # xmm2 \leftarrow_{128} right[i+0...i+3]
  psubd %xmm2, %xmm0
                                     # %xmm0 +128 %xmm0 -128 %xmm2
 movdgu 16(%rsi,%rax,4), %xmm2
                                      # %xmm2 ←128 right[i+4...i+7]
  psubd %xmm2, %xmm1
                                      # %xmm1 + 128 %xmm1 - 128 %xmm2
 movdqu %xmm0, (%rdx,%rax,4)
                                     # result[i+0...i+3] ←<sub>128</sub> %xmm0
                                                                                                loop #n
 movdgu %xmm1, 16(%rdx,%rax,4)
                                     # result[i+4...i+7] ←128 %xmm1
 movdgu 32(%rdi,%rax,4), %xmm0
                                     # %xmm0 ←128 left[i+8 ...i+11]
 movdgu 48(%rdi,%rax,4), %xmm1
                                     # xmm1 \leftarrow_{128} left[i+12...i+15]
                                                                                                loop #n+1
 movdqu 32(%rsi,%rax,4), %xmm2
                                     # xmm2 \leftarrow_{128} right[i+8...i+11]
  psubd %xmm2, %xmm0
                                      # %xmm0 +128 %xmm0 -128 %xmm2
 movdqu 48(%rsi,%rax,4), %xmm2
                                     # %xmm2 ←<sub>128</sub> right[i+12...i+15]
  psubd %xmm2, %xmm1
                                      # %xmm1 + 128 %xmm1 - 128 %xmm2
 movdgu %xmm0, 32(%rdx,%rax,4)
                                     # result[i+8 ...i+11] ←<sub>128</sub> %xmm0
                                     # result[i+12...i+15] +128 %xmm1
 movdgu %xmm1, 48(%rdx,%rax,4)
  addq $16, %rax
                                      # \int 1024 / 16 = 64 iterations
  cmpg $1024, %rax
                                     # exit if %rax = 1024
 ine loop
                                      # (non-vectorized code not shown)
```