Push vs. Pull-Based Loop Fusion in Query Engines

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Introduction

- DBMSes are essential components of software systems
- Persistence layer
- Expose a declarative language (SQL) to users

Query Processing



Physical Query Plan

- Join has different implementations
 - Hash Join
 - Nested Loop Join
 - Sort Merge Join
- Each one is appropriate for a particular scenario
- Physical Query Plan
 - Annotated Relational Algebra
 - Concrete implementation choice for each operator
 - Can improve the execution time of a query from 1 year to 0.1 seconds

Query Interpretation

- Runtime library
- Iterator Model (pull based)
- Used in mainstream DBMSes for a long time

Dominating cost is I/O

Numbers Everyone Should Know

L1 cache reference	0	.5 ns
Branch mispredict	5	ns
L2 cache reference	7	ns
Mutex lock/unlock	25	ns
Main memory reference	100	ns
Compress 1K bytes with Zippy	3,000	ns
Send 2K bytes over 1 Gbps network	20,000	ns
Read 1 MB sequentially from memory	250,000	ns
Round trip within same datacenter	500,000	ns
Disk seek	10,000,000	ns
Read 1 MB sequentially from disk	20,000,000	ns
Send packet CA->Netherlands->CA	150,000,000	ns



In-Memory Databases

- Modern Hardware Technology
- Servers with 1TB of RAM
- The whole database can fit in the RAM

Code layout is important

Why Query Compilation?



Interpreted Query

- ✓ Good out-of-core data locality
- X Virtual function calls
 - X Bad branch prediction & cache locality

Compiled Query

✓ Good branch prediction
 & cache locality

Push vs. Pull



Push vs. Pull

```
var sum = 0.0
var sum = 0.0
                              var index = 0
var index = 0
                              while(true) {
                                var rec = null
                                do {
                                 if(index < R.length) {</pre>
while(index < R.length) {</pre>
                                   rec = R(index)
 val rec = R(index)
                                   index += 1
 index += 1
                                 } else {
                                   rec = null
                                 }
                                } while(rec != null && !(rec.A < 10))</pre>
 if(rec.A < 10)
                                if(rec == null) break
                                else sum += rec.B
   sum += rec.B
}
                              }
return sum
                              return sum
```

CFG of Push vs. Pull



Pull Engine produces a more complicated CFG

Simplifying CFG



Optimizing Compilers successfully simplify CFG

Push Engine Issues

- Hard to handle
 - Merge & Zip-like operator
 - Limit operator
- Solutions
 - Give up & materialize
 - Ad-hoc fused version of operators
 - Rely on hacky mechanisms
 - Makes the interface more complicated

Pipelining in SQL = Fusion in collections

R

SELECT SUM(R.B) FROM R WHERE R.A < 10

```
var sum = 0.0
var index = 0
while(index < R.length) {
   val rec = R(index)
   index += 1
   if(rec.A < 10) sum += rec.B
}
return sum</pre>
```

Pipelining/Fusion History



Push

```
class ProjectOp[R, P](f: R => P) {
                                          def map[S](f: R => S) = build { k =>
 def consume(e: R): Unit =
                                            for(e <- this)</pre>
    dest.consume(f(e))
                                               k(f(e))
                                          }
}
class SelectOp[R](p: R => Boolean) {
                                          def filter(p: R => Boolean) = build { k =>
 def consume(e: R): Unit =
                                            for(e <- this)</pre>
   if(p(e))
                                             if(p(e))
       dest.consume(e)
                                                k(e)
                                          }
}
```

Pull

```
class SelectOp[R](p: R => Boolean) {
                                        def filter(p: R=>Boolean) = destroy { n =>
 def next(): R = {
                                          generate { () =>
   var elem: R = null
                                           var elem: R = null
   do {
                                           do {
                                              elem = n()
      elem = source.next()
   } while (elem != null && !p(elem))
                                           } while(elem != null && !p(elem))
  elem
                                           elem
} }
                                        } }
class ProjectOp[R, P](f: R => P) {
                                        def map[P](p: R => P) = destroy { n =>
 def next(): P = {
                                          generate { () =>
   val elem = source.next()
                                           val elem = n()
   if(elem == null) null
                                           if(elem == null) null
                                           else f(elem)
   else f(elem)
} }
                                        } }
```

Fusion/Pipelining Correspondence

- Fold Fusion = Push Engine
- Unfold Fusion = Pull Engine



Stream Fusion = Stream-Fusion Engine

Stream-Fusion Engine

- A pipelined query engine
- Inspired by the Stream Fusion approach developed for functional collections
- Combines the benefits of pull and pushbased engines

Stream Fusion

```
class SelectOp[R](p: R => Boolean) {
  def stream(): Step[R] = {
    source.stream().filter(p)
  }
  }
class ProjectOp[R, P](f: R => P) {
    def stream(): Step[P] = {
        source.stream().map(f)
  }
}
```

```
def filter(p: R => Boolean) = {
    unstream { () =>
        stream().filter(p)
} 
def map[P](f: R => P) = {
    unstream { () =>
        stream().map(f)
}
```

Intermediate Step Objects

- Stream-fusion creates intermediate step objects:
 - Yield
 - Skip
 - Done
- Two solutions to remove their allocation
 - Scalar replacement
 - Church encoding of the sum type

Type of engine	Run time (ms)
Pull (Interpreted)	3,486
Pull (Naïve)	2,405
Pull (Inline-Friendly)	2,165
Stream (Scalar replacement for Step objects)	2,447
Stream (Visitor model for Step objects)	2,217
Stream (No removal of Step objects)	6,886

Micro-Benchmark Results



Micro-Benchmark Results (Cont.)



Experimental Results



In most cases all show similar performance

Conclusions

- Pipelining == Fusion
- •DB ⇔ PL
- Many techniques can be exchanged
 - Vectorization == Generalized Stream-Fusion
 - Query optimizers
 - Column stores

Thank You!