Scalarization of Index Vectors in Compiled APL

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Abstract

High-performance for array languages offers several unique challenges to the compiler writer, including fusion of loops over large arrays, detection and elimination of scalars as arbitrary arrays, and eliminating or minimizing the run-time creation of index vectors.

We introduce one of those challenges in the context of SAC, a functional array languge, and give preliminary results on the performance of a compiler that eliminates index vectors by scalarizing them within the optimization cycle.

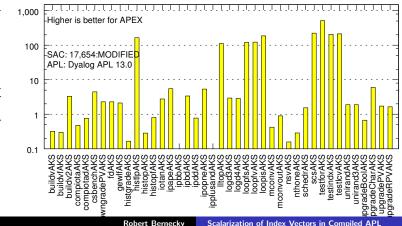
- How much faster is compiled APL than interpreted APL?
- The answer is **NOT** a scalar.

- Dyalog APL 13.0 vs. APEX/SAC
- The current SAC compiler
 - a functional array language
 - data-parallel nested loops: With-Loop
 - array-based optimizations
 - functional loops and conditionals as functions
- Goal: Compiled APL performance competitive with hand-coded C

Some Reasons Why APL is Slow

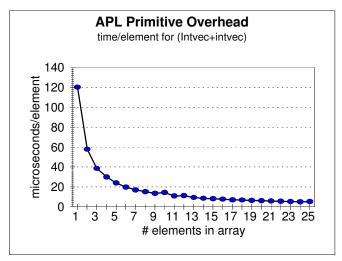
- Fixed per-primitive overheads: Syntax analysis, conf checks, fn dispatch, mem mgmt
- Variable per-primitive overheads
- Index vector materialization

APL vs. APEX CPU Time Performance (2,011-09-30)



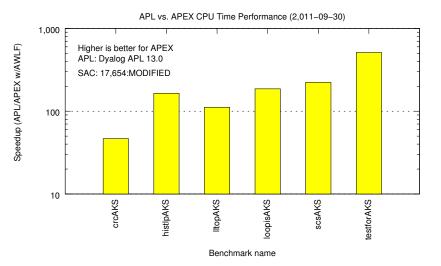
Speedup (APL/APEX w/AWLF)

Why APL is Slow: Fixed Per-Primitive Overheads



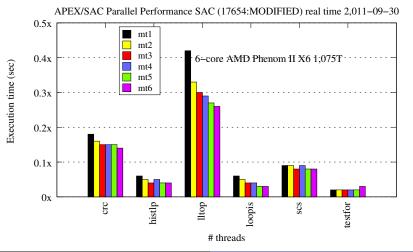
Who suffers? Apps dominated by operations on scalars: CRC, loopy histograms, dynamic programming, RNG

Why APL is Slow: Fixed Per-Primitive Overheads



Why APL is Slow: Fixed Per-Primitive Overheads

- Scalar-dominated apps have good serial speedup...
- but poor parallel speedup

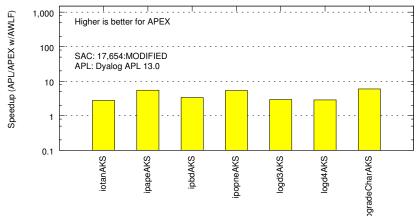


Robert Bernecky Scalarization of Index Vectors in Compiled APL

Why is APL Slow? Variable Per-Primitive Overheads

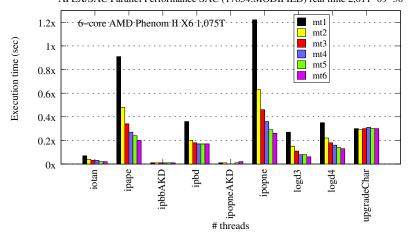
- Naive execution: Limited fn composition, e.g. sum(iota(N))
- Array-valued intermediate results: memory madness
- Who suffers? Apps dominated by operations on large arrays: Signal processing, convolution, normal move-out

APL vs. APEX CPU Time Performance (2,011-09-30)



Why is APL Slow? Variable Per-Primitive Overheads

 Who suffers? Apps dominated by operations on large arrays: Signal processing, convolution, normal move-out APEX/SAC Parallel Performance SAC (17654:MODIFIED) real time 2,011–09–30



- Mike Jenkins' matrix divide model a[;i,pi] = a[;pi,i]
- [i,pi] and [pi,i] are materialized index vectors
- A few simple changes to scalarize index vectors: tmp = a[;i] a[;i] = a[;pi] a[;pi] = tmp
- Matrix divide model now runs twice as fast!

Why Materialized Index Vectors are Expensive

Materialization of [i,pi] and [pi,i]:

 (* for indexing part)
 *Increment reference counts on i and pi Allocate 2-element temp vector
 Initialize temp vector descriptor
 Initialize temp vector elements
 *Perform indexing
 Deallocate 2-element temp vector
 *Decrement reference counts on i and pi

- Who suffers? Apps using explicit array indexing
- e.g., many apps dominated by indexed assign
- Who suffers? Matrix divide, compress, deal, dynamic programming
- Who suffers? Inner products that use the CDC STAR-100 algorithm
- 800x800 ipplusandAKD CPU time: 45 minutes!

Eliminating Materialized Index Vectors With IVE

- IFL2006 paper: Index Vector Elimination (IVE) Bernecky, Grelck, Herhut, Scholz, Trojahner, and Schafarenko
- Post-optimization transformation
- Start with:

IV = [i, j, k]

```
z = sel( IV, M)
```

IVE: Replace: z = M[IV] by: IV = [i, j, k]

offset = vect2offset(shape(M), IV)

```
z = idxsel( offset, M)
```

- Implication: IV is a materialized index vector!
- Implication: offset calculation may be liftable (LIR)

Eliminating Materialized Index Vectors With IVE

```
IVE: If IV can be represented as scalars, eliminate vect2offset. Before:
IV = [ i, j, k]
offset = vect2offset( shape(M), IV)
z = idxsel( offset, M)
After:
IV = [ i, j, k]
offset = idxs2offset( shape(M), i, j, k)
z = idxsel( offset, M)
```

IV is now dead code, and can be eliminated!

Eliminating Materialized Index Vectors With IVE

```
IVE: If IV is NOT scalars, scalarize it. Before:
offset = vect2offset( shape(M), IV)
z = idxsel( offset, M)
After:
I = IV[0]
J = IV[1]
K = IV[2]
JV = [ I, J, K]
offset = vect2offset( shape(M), JV)
```

- IV is now dead code, and can be eliminated!
- Earlier substitution by idxs2offset now feasible.
- Unfortunately...

Dueling Optimizations: IVE vs. LIR

- IVE introduces JV = [I, J, K]
- Loop-Invariant Removal (LIR) lifts I, J, K, JV out of the function
- Constant Folding (CF) replaces JV = [I, J, K] by JV = IV
- Common-subexpression elimination (CSE) replaces JV by IV
- This is where we came in!
- So, what can we do?
- A kludged LIR to deal with this was deemed tasteless

- Another approach: move IVE into the optimization cycle
- Having IVE in opt cycle enables other optimizations!
- But, it also breaks many optimizations
- ► To prevent dueling, we must scalarize all index vector ops!
- *e.g.*, unroll IV+1, guard and extrema functions..
- *e.g.*, extend existing optimizations, such as CF
- The Good News: I had already scalarized many index vectors for Algebraic With-Loop Folding!

- Moving IVE into optimization cycle worked, sort of:
- The Good News: ipplusandAKD CPU time: 8 seconds, instead of 45 minutes
- The Bad News: New primitives in AST broke some optimizations!
- More Bad News: I am still fixing them!
- ► The Good News: Performance is improving daily
- More Good News: Many opportunities exist for further optimization
- Even More Good News: More parallelism is coming

- Array sizes affect performance
- Iteration counts affect performance
- Indexed assigns affect performance
- Characterize your application, then we can provide an answer.